75

Un upper bound for Kullback-Leibler divergence with a small number of outliers

Alexander Gofman¹ and Mark Kelbert^{2,*}

Received November 24, 2011; accepted October 22, 2012

Abstract. We establish a new upper bound for the Kullback-Leibler divergence of two discrete probability distributions which are close in a sense that typically the ratio of probabilities is nearly one and the number of outliers is small.

AMS subject classifications: 94A17, 62B10

Key words: Kullback-Leibler divergence, relative entropy, mutual information, information inequality

1. Introduction

Let $\mathbb{P}=(p_i)$ and $\mathbb{Q}=(q_i)$ be two probability distributions on the set $\mathcal{R}=\{1,2,\ldots,r\}$ and

$$D(\mathbb{P} \parallel \mathbb{Q}) = \sum_{i=1}^{r} p_i \ln \frac{p_i}{q_i}$$
 (1)

the Kullback-Leibler divergence between these distributions (cf. [1, 3]). The problem of obtaining good upper and lower bounds for the relative entropy attracts considerable interest in information theory (see [1-4]). We present here two famous upper bounds (see proofs in [2]):

$$D(\mathbb{P} \| \mathbb{Q}) \le \min \left[\sum_{i=1}^{r} \frac{p_i^2}{q_i} - 1, \sum_{i=1}^{r} \sqrt{\frac{p_i}{q_i}} |p_i - q_i| \right].$$
 (2)

In the case $|\frac{p_i}{q_i} - 1| \le \epsilon$ for all i the first bound gives an estimate of order $O(\epsilon)$. In the second bound the distributions should be close in the \mathbf{L}_1 sense. We present here a new upper bound for $D(\mathbb{P} \parallel \mathbb{Q})$ of the first type. However, we achieve the order of approximation $O(\epsilon^2)$ and also allow a possibility of outliers for \mathbb{P} and \mathbb{Q} with a small total mass. We say that $(\mathbb{P}, \mathbb{Q}) \in \mathcal{C}(\epsilon, \delta, c, d)$ if there exists a partition $\mathcal{R} = \mathcal{R}_1 \cup \mathcal{R}_2$ where \mathcal{R}_1 is called a regular set and \mathcal{R}_2 is the set of outliers with the following properties. Assume that for some $\epsilon, \delta > 0$ and some constant $0 < c < 1 < d < \infty$

¹ Faculty of Economics, Moscow Economics National Research University, 20 Myasnitskaya str., Moscow, 101 000, RF

² Department of Mathematics, Swansea University, Singleton Park, Swansea, SA2 8PP, UK

^{*}Corresponding author. *Email addresses:* agofman@rambler.ru (A. Gofman), m.kelbert@swansea.ac.uk (M. Kelbert)

$$\left| \frac{q_i}{p_i} - 1 \right| \le \epsilon \ \forall i \in \mathcal{R}_1, \sum_{j \in \mathcal{R}_2} q_j = \delta, c \le \frac{q_j}{p_j} \le d \ \forall j \in \mathcal{R}_2.$$
 (3)

The main result is the upper bound of the Kullback-Leibler divergence for small enough $\epsilon, \delta > 0$:

Theorem 1. For any $(\mathbb{P}, \mathbb{Q}) \in \mathcal{C}(\epsilon, \delta, c, d)$

$$0 \le D(\mathbb{P} \parallel \mathbb{Q}) \le \phi(\epsilon) + \psi(\epsilon, \delta), \tag{4}$$

where $\phi(\epsilon) = \frac{1}{2}\epsilon^2 + 0(\epsilon^3)$, $\psi(\epsilon, \delta) \leq C\delta$ for ϵ , δ small enough, and $C = 2d - c - 1 + \ln \frac{1}{c}$.

Explicit formulas for the functions ϕ and ψ are presented in (12 a, b).

Corollary 1. Suppose that the joint distribution $P_{X,Y}$ on $\mathbb{R}^{\times 2}$ of random variables X and Y satisfies for all i, j the inequality

$$|p_{X,Y}(i,j)/p_X(i)p_Y(j)-1| \le \epsilon.$$

Then the mutual information between X and Y

$$I(X:Y) = D(p_{X|Y} \parallel p_X \times p_Y) < \phi(\epsilon).$$

2. Bound on the ratio of arithmetic and geometric means

Let $0 < c < a < b < d < \infty$, $x_i \in [a,b], i = 1,\ldots,r_1, y_j \in [c,a) \cup (b,d], j = 1,\ldots,r_2$ and $\sum_{j=1}^{r_2} y_j = \delta, r = r_1 + r_2$. Following the strategy used in [5], write $x_i = \lambda_i a + (1 - \lambda_i)b$ and $y_j = \mu_j c + (1 - \mu_j)d$ with $0 \le \lambda_i, \mu_j \le 1$. Then for any distribution $\mathbb P$ on $\mathcal R$ and any convex function f on [c,d]

$$0 \le \sum_{i=1}^{r_1+r_2} p_i f(t_i) - f(\sum_{i=1}^{r_1+r_2} p_i t_i) = S - T$$
 (5)

with $t_i \in \{x_i\} \cup \{y_i\}$. Let us estimate the first sum as follows

$$S \leq \sum_{i=1}^{r_1} p_i (\lambda_i f(a) + (1 - \lambda_i) f(b)) + \sum_{i=1}^{r_2} p_i (\mu_i f(c) + (1 - \mu_i) f(d))$$
$$= (f(a) - f(b)) p_{\lambda} + f(b) (1 - \delta) + (f(c) - f(d)) p_{\mu} + f(d) \delta$$

with
$$p_{\lambda} = \sum_{i=1}^{r_1} p_i \lambda_i, p_{\mu} = \sum_{j=1}^{r_2} p_j \mu_j \le \delta.$$

If we select the convex function as $f(x) = -\ln x$ inequality (5) may be considered as a logarithmic form of the inequality for the ratio of arithmetic and geometric means. We proceed by specifying the second term

$$-T = \ln\left((a-b)p_{\lambda} + b(1-\delta) + (c-d)p_{\mu} + d\delta\right)$$

and using the Taylor expansion

$$-T = \ln\left((a-b)p_{\lambda} + b(1-\delta) + d\delta\right) + \frac{1}{\xi}(c-d)p_{\mu}$$

with $(a-b)p_{\lambda} + b(1-\delta) + d\delta + (c-d)p_{\mu} < \xi < (a-b)p_{\lambda} + b(1-\delta) + d\delta$. So, the RHS of (5) for the function $f(x) = -\ln x$ is estimated as follows:

$$S - T \le \left(\ln \frac{b}{a}\right)p_{\lambda} + \left(\ln \frac{d}{c}\right)p_{\mu} + C_1 + \ln\left[(a-b)p_{\lambda} + b(1-\delta) + d\delta\right] - \frac{d-c}{\xi}p_{\mu} \quad (6)$$

with $C_1 = -(1 - \delta) \ln b - \delta \ln d$. An upper bound will be obtained if we estimate ξ from above $\xi < b(1 - \delta) + d\delta$ as c < d. This result provides an upper bound which is a linear function of p_{μ} :

$$S - T \le \left(\ln\frac{b}{a}\right)p_{\lambda} + K(d, c, b, \delta)p_{\mu} + C_1 + \ln\left[(a - b)p_{\lambda} + b(1 - \delta) + d\delta\right]. \tag{7}$$

In case the coefficient

$$K = K(d, c, b, \delta) = \ln \frac{d}{c} - \frac{d - c}{b(1 - \delta) + d\delta} > 0,$$

an upper bound is obtained by selecting $p_{\mu} = \delta$. In the case K < 0 we select $p_{\mu} = 0$. In any case, the term proportional to p_{μ} does not depend on p_{λ} and the value p_{λ}^* maximizing the RHS in (7) is the solution of equation

$$\ln b - \ln a + \frac{a - b}{(a - b)p_{\lambda} + b(1 - \delta) + d\delta} = 0, \text{ i.e. } p_{\lambda}^* = \frac{b(1 - \delta) + d\delta}{b - a} - \frac{1}{\ln b - \ln a}.$$
 (8)

It is straightforward to check that the second derivative is negative at p_{λ}^* . Next, note that

$$p_{\lambda}^* > \frac{b}{b-a} - \frac{1}{\ln b - \ln a} > 0$$

and $p_{\lambda}^* < 1 - \delta$ for $\delta < \delta^* = \frac{1}{d-a} \left(\frac{b-a}{\ln b - \ln a} - a \right)$. Let $L(a,b) = \frac{b-a}{\ln b - \ln a}$. Combining all results together we obtain that

$$S - T \le \ln \Lambda(a, b) + \delta \Gamma(a, b, c, d, \delta), \tag{9}$$

where

$$\ln \Lambda(a,b) = \frac{b}{L(a,b)} - 1 - \ln b + \ln(L(a,b)). \tag{10}$$

The function $\Lambda(a,b)$ in first term may be represented as follows (cf. [5]):

$$\Lambda(a,b) = abL(a,b)/G^{2}(a,b), G(a,b) = \frac{1}{e} (b^{b}/a^{a})^{1/(b-a)}.$$

in the case $K(d, c, b, \delta) > 0$ we have

$$\Gamma(a, b, c, d, \delta) = \frac{(d-a)\ln b - (d-b)\ln a}{b-a} - \ln c + \frac{d-c}{b+(d-b)\delta}$$
(11a)

and in the case $K(d, c, b, \delta) \leq 0$

$$\Gamma(a,b,c,d,\delta) = \frac{(d-a)\ln b - (d-b)\ln a}{b-a} - \ln c.$$
(11b)

Observe that $\Gamma(a, b, c, d, \delta) > 0$ in both cases.

3. Proof of the Theorem 1

Let us apply the results of the previous section with $t_i = \frac{q_i}{p_i}$, $a = 1 - \epsilon$, $b = 1 + \epsilon$. By straightforward calculations the expression in (10) admits the expansion

$$\ln \Lambda(1-\epsilon,1+\epsilon) = \epsilon + \frac{1}{3}\epsilon^2 - \epsilon + \frac{1}{2}\epsilon^2 - \frac{1}{3}\epsilon^2 + \dots = \frac{1}{2}\epsilon^2 + \dots$$

For ϵ small enough both expressions in (11a,b) are bounded by $2d-c-1+\ln\frac{1}{c}$ by inspection. Now the Theorem follows from (9) with

$$\phi(\epsilon) = (1+\epsilon)F(\epsilon)^{-1} - 1 - \ln(1+\epsilon) - \ln F(\epsilon), \quad F(\epsilon) = \left(\ln \frac{1+\epsilon}{1-\epsilon}\right)/2\epsilon$$
 (12a)

and

$$\psi(\epsilon, \delta) = \delta \left[(d - 1 + \epsilon) F(\epsilon) - \ln c + \frac{d - c}{1 + (d - 1 - \epsilon) \delta} \right].$$
 (12b)

References

- T. M. COVER, J. M. THOMAS, Elements of Information Theory, Wiley, New York, 2006.
- [2] S. S. DRAGOMIR, V. GLUŠČEVIĆ, New estimates of the Kullback-Leibler distance and applications, in: Inequality Theory and Applications Vol. I, (Y. J. Cho, J. K. Kim and S. S. Dragomir, Eds.), Nova Sci. Publ., Huntington, NY, 2001, 123–137.
- [3] M. Kelbert, Y. Suhov, Information Theory and Coding by Example, Cambridge Univ. Press, Cambridge, 2013.
- [4] A. SAYYAREH, A new upper bound for Kullback-Leibler divergence, Appl. Math. Sci. 67(2011), 3303–3317.
- [5] S. Simic, Best possible global bounds for Jensen functional, AMS Proceedings 138(2010), 2457–2462.