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GENERALIZED RELATIVE INFORMATION AND INFORMATION INEQUALITIES

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Abstract

In this paper, we have obtained bounds on Csiszár's *f-divergence* in terms of *relative information of type s* using Dragomir's [9] approach. The results obtained in particular lead us to bounds in terms of χ^2 –Divergence, Kullback-Leibler's *relative information* and Hellinger's *discrimination*.

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Key words: Relative information; Csiszár's f-divergence; χ^2 -divergence; Hellinger's discrimination; Relative information of type s; Information inequalities.

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1. Introduction

Let

$$\Delta_n = \left\{ P = (p_1, p_2, \dots, p_n) \middle| p_i > 0, \sum_{i=1}^n p_i = 1 \right\}, \quad n \ge 2,$$

be the set of complete finite discrete probability distributions.

The Kullback Leibler's [13] relative information is given by

(1.1)
$$K(P||Q) = \sum_{i=1}^{n} p_i \ln\left(\frac{p_i}{q_i}\right),$$

for all $P, Q \in \Delta_n$.

In Δ_n , we have taken all $p_i > 0$. If we take $p_i \geq 0, \forall i = 1, 2, ..., n$, then in this case we have to suppose that $0 \ln 0 = 0 \ln \left(\frac{0}{0}\right) = 0$. From the *information theoretic* point of view we generally take all the logarithms with base 2, but here we have taken only natural logarithms.

We observe that the measure (1.1) is not symmetric in P and Q. Its symmetric version, famous as J-divergence (Jeffreys [12]; Kullback and Leiber [13]), is given by

(1.2)
$$J(P||Q) = K(P||Q) + K(Q||P) = \sum_{i=1}^{n} (p_i - q_i) \ln\left(\frac{p_i}{q_i}\right).$$

Let us consider the one parametric generalization of the measure (1.1), called *relative information of type s* given by

(1.3)
$$K_s(P||Q) = [s(s-1)]^{-1} \left[\sum_{i=1}^n p_i^s q_i^{1-s} - 1 \right], \quad s \neq 0, 1.$$



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In this case we have the following limiting cases

$$\lim_{s \to 1} K_s(P||Q) = K(P||Q),$$

and

$$\lim_{s\to 0} K_s(P||Q) = K(Q||P).$$

The expression (1.3) has been studied by Vajda [22]. Previous to it many authors studied its characterizations and applications (ref. Taneja [20] and on line book Taneja [21]).

We have some interesting particular cases of the measure (1.3).

(i) When $s = \frac{1}{2}$, we have

$$K_{1/2}(P||Q) = 4[1 - B(P||Q)] = 4h(P||Q)$$

where

(1.4)
$$B(P||Q) = \sum_{i=1}^{n} \sqrt{p_i q_i},$$

is the famous as Bhattacharya's [1] distance, and

(1.5)
$$h(P||Q) = \frac{1}{2} \sum_{i=1}^{n} (\sqrt{p_i} - \sqrt{q_i})^2,$$

is famous as Hellinger's [11] discrimination.



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(ii) When s = 2, we have

$$K_2(P||Q) = \frac{1}{2}\chi^2(P||Q),$$

where

(1.6)
$$\chi^{2}(P||Q) = \sum_{i=1}^{n} \frac{(p_{i} - q_{i})^{2}}{q_{i}} = \sum_{i=1}^{n} \frac{p_{i}^{2}}{q_{i}} - 1,$$

is the χ^2 -divergence (Pearson [16]).

(iii) When s = -1, we have

$$K_{-1}(P||Q) = \frac{1}{2}\chi^2(Q||P),$$

where

(1.7)
$$\chi^2(Q||P) = \sum_{i=1}^n \frac{(p_i - q_i)^2}{p_i} = \sum_{i=1}^n \frac{q_i^2}{p_i} - 1.$$

For simplicity, let us write the measures (1.3) in the unified way:

(1.8)
$$\Phi_s(P||Q) = \begin{cases} K_s(P||Q), & s \neq 0, 1, \\ K(Q||P), & s = 0, \\ K(P||Q), & s = 1. \end{cases}$$



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Summarizing, we have the following particular cases of the measures (1.8):

(i)
$$\Phi_{-1}(P||Q) = \frac{1}{2}\chi^2(Q||P)$$
.

(ii)
$$\Phi_0(P||Q) = K(Q||P)$$
.

(iii)
$$\Phi_{1/2}(P||Q) = 4[1 - B(P||Q)] = 4h(P||Q).$$

(iv)
$$\Phi_1(P||Q) = K(P||Q)$$
.

(v)
$$\Phi_2(P||Q) = \frac{1}{2}\chi^2(P||Q)$$
.



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2. Csiszár's f-Divergence and Information Bounds

Given a convex function $f:[0,\infty)\to\mathbb{R}$, the f-divergence measure introduced by Csiszár [4] is given by

(2.1)
$$C_f(p,q) = \sum_{i=1}^n q_i f\left(\frac{p_i}{q_i}\right),$$

where $p, q \in \mathbb{R}^n_+$.

The following two theorems can be seen in Csiszár and Körner [5].

Theorem 2.1. (Joint convexity). If $f : [0, \infty) \to \mathbb{R}$ be convex, then $C_f(p, q)$ is jointly convex in p and q, where $p, q \in \mathbb{R}^n_+$.

Theorem 2.2. (Jensen's inequality). Let $f:[0,\infty)\to\mathbb{R}$ be a convex function. Then for any $p,q\in\mathbb{R}^n_+$, with $P_n=\sum_{i=1}^n p_i>0, Q_n=\sum_{i=1}^n p_i>0$, we have the inequality

$$C_f(p,q) \ge Q_n f\left(\frac{P_n}{Q_n}\right).$$

The equality sign holds for strictly convex functions iff

$$\frac{p_1}{q_i} = \frac{p_2}{q_2} = \dots = \frac{p_n}{q_n}.$$

In particular, for all $P, Q \in \Delta_n$, we have

$$C_f(P||Q) \ge f(1),$$

with equality iff P = Q.

In view of Theorems 2.1 and 2.2, we have the following result.



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Result 1. For all $P, Q \in \Delta_n$, we have

- (i) $\Phi_s(P||Q) \geq 0$ for any $s \in \mathbb{R}$, with equality iff P = Q.
- (ii) $\Phi_s(P||Q)$ is convex function of the pair of distributions $(P,Q) \in \Delta_n \times \Delta_n$ and for any $s \in \mathbb{R}$.

Proof. Take

(2.2)
$$\phi_s(u) = \begin{cases} [s(s-1)]^{-1} [u^s - 1 - s(u-1)], & s \neq 0, 1; \\ u - 1 - \ln u, & s = 0; \\ 1 - u + u \ln u, & s = 1 \end{cases}$$

for all u > 0 in (2.1), we have

$$C_f(P||Q) = \Phi_s(P||Q) = \begin{cases} K_s(P||Q), & s \neq 0, 1; \\ K(Q||P), & s = 0; \\ K(P||Q), & s = 1. \end{cases}$$

Moreover,

(2.3)
$$\phi'_s(u) = \begin{cases} (s-1)^{-1} (u^{s-1} - 1), & s \neq 0, 1; \\ 1 - u^{-1}, & s = 0; \\ \ln u, & s = 1 \end{cases}$$



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(2.4)
$$\phi_s''(u) = \begin{cases} u^{s-2}, & s \neq 0, 1; \\ u^{-2}, & s = 0; \\ u^{-1}, & s = 1. \end{cases}$$

Thus we have $\phi_s''(u) > 0$ for all u > 0, and hence, $\phi_s(u)$ is strictly convex for all u > 0. Also, we have $\phi_s(1) = 0$. In view of Theorems 2.1 and 2.2 we have the proof of parts (i) and (ii) respectively.

For some studies on the measure (2.2) refer to Liese and Vajda [15], Österreicher [17] and Cerone et al. [3].

The following theorem summarizes some of the results studies by Dragomir [7], [8]. For simplicity we have taken f(1) = 0 and $P, Q \in \Delta_n$.

Theorem 2.3. Let $f: \mathbb{R}_+ \to \mathbb{R}$ be differentiable convex and normalized i.e., f(1) = 0. If $P, Q \in \Delta_n$ are such that

$$0 < r \le \frac{p_i}{q_i} \le R < \infty, \quad \forall i \in \{1, 2, \dots, n\},$$

for some r and R with $0 < r \le 1 \le R < \infty$, then we have the following inequalities:

(2.5)
$$0 \le C_f(P||Q) \le \frac{1}{4} (R-r) \left(f'(R) - f'(r) \right),$$

$$(2.6) 0 \le C_f(P||Q) \le \beta_f(r,R),$$



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(2.7)
$$0 \leq \beta_{f}(r,R) - C_{f}(P||Q)$$

$$\leq \frac{f'(R) - f'(r)}{R - r} \left[(R - 1)(1 - r) - \chi^{2}(P||Q) \right]$$

$$\leq \frac{1}{4} (R - r) \left(f'(R) - f'(r) \right),$$

where

(2.8)
$$\beta_f(r,R) = \frac{(R-1)f(r) + (1-r)f(R)}{R-r},$$

and $\chi^2(P||Q)$ and $C_f(P||Q)$ are as given by (1.6) and (2.1) respectively.

In view of above theorem, we have the following result.

Result 2. Let $P, Q \in \Delta_n$ and $s \in \mathbb{R}$. If there exists r, R such that

$$0 < r \le \frac{p_i}{q_i} \le R < \infty, \quad \forall i \in \{1, 2, \dots, n\},$$

with $0 < r \le 1 \le R < \infty$, then we have

(2.9)
$$0 \le \Phi_s(P||Q) \le \mu_s(r, R),$$

$$(2.10) 0 \le \Phi_s(P||Q) \le \phi_s(r,R),$$



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(2.11)
$$0 \le \phi_s(r, R) - \Phi_s(P||Q) \\ \le k_s(r, R) \left[(R - 1)(1 - r) - \chi^2(P||Q) \right] \\ \le \mu_s(r, R),$$

where

(2.12)
$$\mu_s(r,R) = \begin{cases} \frac{1}{4} \frac{(R-r)(R^{s-1}-r^{s-1})}{(s-1)}, & s \neq 1; \\ \frac{1}{4}(R-r)\ln\left(\frac{R}{r}\right), & s = 1 \end{cases}$$

(2.13)
$$\phi_{s}(r,R) = \frac{(R-1)\phi_{s}(r) + (1-r)\phi_{s}(R)}{R-r}$$

$$= \begin{cases} \frac{(R-1)(r^{s}-1) + (1-r)(R^{s}-1)}{(R-r)s(s-1)}, & s \neq 0, 1; \\ \frac{(R-1)\ln\frac{1}{r} + (1-r)\ln\frac{1}{R}}{(R-r)}, & s = 0; \\ \frac{(R-1)r\ln r + (1-r)R\ln R}{(R-r)}, & s = 1, \end{cases}$$

and

(2.14)
$$k_s(r,R) = \frac{\phi_s'(R) - \phi_s'(r)}{R - r} = \begin{cases} \frac{R^{s-1} - r^{s-1}}{(R - r)(s - 1)}, & s \neq 1; \\ \frac{\ln R - \ln r}{R - r}, & s = 1. \end{cases}$$



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Proof. The above result follows immediately from Theorem 2.3, by taking $f(u) = \phi_s(u)$, where $\phi_s(u)$ is as given by (2.2), then in this case we have $C_f(P||Q) = \Phi_s(P||Q)$.

Moreover,

$$\mu_s(r,R) = \frac{1}{4}(R-r)^2 k_s(r,R),$$

where

$$k_s(r,R) = \begin{cases} \left[L_{s-2}(r,R)\right]^{s-2}, & s \neq 1; \\ \left[L_{-1}(r,R)\right]^{-1} & s = 1, \end{cases}$$

and $L_p(a,b)$ is the famous (ref. Bullen, Mitrinović and Vasić [2]) *p-logarithmic* mean given by

$$L_p(a,b) = \begin{cases} \left[\frac{b^{p+1} - a^{p+1}}{(p+1)(b-a)} \right]^{\frac{1}{p}}, & p \neq -1, 0; \\ \frac{b-a}{\ln b - \ln a}, & p = -1; \\ \frac{1}{e} \left[\frac{b^b}{a^a} \right]^{\frac{1}{b-a}}, & p = 0, \end{cases}$$

for all $p \in \mathbb{R}$, $a, b \in \mathbb{R}_+$, $a \neq b$.

We have the following corollaries as particular cases of Result 2.

Corollary 2.4. *Under the conditions of Result 2, we have*

(2.15)
$$0 \le \chi^2(Q||P) \le \frac{1}{4}(R+r) \left(\frac{R-r}{rR}\right)^2,$$

(2.16)
$$0 \le K(Q||P) \le \frac{(R-r)^2}{4Rr},$$



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$$(2.17) 0 \le K(P||Q) \le \frac{1}{4}(R-r)\ln\left(\frac{R}{r}\right),$$

(2.18)
$$0 \le h(P||Q) \le \frac{(R-r)\left(\sqrt{R} - \sqrt{r}\right)}{8\sqrt{Rr}}$$

(2.19)
$$0 \le \chi^2(P||Q) \le \frac{1}{2}(R-r)^2.$$

Proof. (2.15) follows by taking s=-1, (2.16) follows by taking s=0, (2.17) follows by taking s=1, (2.18) follows by taking $s=\frac{1}{2}$ and (2.19) follows by taking s=2 in (2.9).

Corollary 2.5. *Under the conditions of Result 2, we have*

(2.20)
$$0 \le \chi^2(Q||P) \le \frac{(R-1)(1-r)}{rR},$$

(2.21)
$$0 \le K(Q||P) \le \frac{(R-1)\ln\frac{1}{r} + (1-r)\ln\frac{1}{R}}{R-r},$$

(2.22)
$$0 \le K(P||Q) \le \frac{(R-1)r\ln r + (1-r)R\ln R}{R-r},$$

$$(2.23) 0 \le h(P||Q) \le \frac{\left(\sqrt{R} - 1\right)(1 - \sqrt{r})}{\sqrt{R} + \sqrt{r}}$$

and

$$(2.24) 0 \le \chi^2(P||Q) \le (R-1)(1-r).$$



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Proof. (2.20) follows by taking s = -1, (2.21) follows by taking s = 0, (2.22) follows by taking s = 1, (2.23) follows by taking $s = \frac{1}{2}$ and (2.24) follows by taking s = 2 in (2.10).

In view of (2.16), (2.17), (2.21) and (2.22), we have the following bounds on *J-divergence*:

$$(2.25) 0 \le J(P||Q) \le \min\left\{t_1(r,R), t_2(r,R)\right\},\,$$

where

$$t_1(r,R) = \frac{1}{4}(R-r)^2 \left[(rR)^{-1} + (L_{-1}(r,R))^{-1} \right]$$

and

$$t_2(r,R) = (R-1)(1-r)(L_{-1}(r,R))^{-1}$$
.

The expression $t_1(r, R)$ is due to (2.16) and (2.17) and the expression $t_2(r, R)$ is due to (2.21) and (2.22).

Corollary 2.6. *Under the conditions of Result 2, we have*

(2.26)
$$0 \le \frac{(R-1)(1-r)}{rR} - \chi^2(Q||P)$$
$$\le \frac{R+r}{(rR)^2} \left[(R-1)(1-r) - \chi^2(P||Q) \right],$$

(2.27)
$$0 \le \frac{(R-1)\ln\frac{1}{r} + (1-r)\ln\frac{1}{R}}{R-r} - K(Q||P)$$
$$\le \frac{1}{rR} \left[(R-1)(1-r) - \chi^2(P||Q) \right],$$



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(2.28)
$$0 \le \frac{(R-1)r\ln r + (1-r)R\ln R}{R-r} - K(P||Q)$$
$$\le \frac{\ln R - \ln r}{R-r} \left[(R-1)(1-r) - \chi^2(P||Q) \right]$$

$$(2.29) 0 \le \frac{\left(\sqrt{R} - 1\right)(1 - \sqrt{r})}{\left(\sqrt{R} + \sqrt{r}\right)} - h(P||Q)$$

$$\le \frac{1}{2\sqrt{rR}\left(\sqrt{R} + \sqrt{r}\right)} \left[(R - 1)(1 - r) - \chi^2(P||Q) \right].$$

Proof. (2.26) follows by taking s=-1, (2.27) follows by taking s=0, (2.28) follows by taking s=1, (2.29) follows by taking $s=\frac{1}{2}$ in (2.11).



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3. Main Results

In this section, we shall present a theorem generalizing the one obtained by Dragomir [9]. The results due to Dragomir [9] are limited only to χ^2 -divergence, while the theorem established here is given in terms of relative information of type s, that in particular lead us to bounds in terms of χ^2 -divergence, Kullback-Leibler's relative information and Hellinger's discrimination.

Theorem 3.1. Let $f: I \subset \mathbb{R}_+ \to \mathbb{R}$ the generating mapping be normalized, i.e., f(1) = 0 and satisfy the assumptions:

- (i) f is twice differentiable on (r, R), where $0 \le r \le 1 \le R \le \infty$;
- (ii) there exists the real constants m, M with m < M such that

$$(3.1) m \le x^{2-s} f''(x) \le M, \quad \forall x \in (r, R), \quad s \in \mathbb{R}.$$

If $P, Q \in \Delta_n$ are discrete probability distributions satisfying the assumption

$$0 < r \le \frac{p_i}{q_i} \le R < \infty,$$

then we have the inequalities:

(3.2)
$$m \left[\phi_s(r,R) - \Phi_s(P||Q) \right] \le \beta_f(r,R) - C_f(P||Q)$$

$$\le M \left[\phi_s(r,R) - \Phi_s(P||Q) \right],$$

where $C_f(P||Q)$, $\Phi_s(P||Q)$, $\beta_f(r,R)$ and $\phi_s(r,R)$ are as given by (2.1), (1.8), (2.8) and (2.13) respectively.



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Proof. Let us consider the functions $F_{m,s}(\cdot)$ and $F_{M,s}(\cdot)$ given by

(3.3)
$$F_{m,s}(u) = f(u) - m\phi_s(u),$$

and

(3.4)
$$F_{M,s}(u) = M\phi_s(u) - f(u),$$

respectively, where m and M are as given by (3.1) and function $\phi_s(\cdot)$ is as given by (2.3).

Since f(u) and $\phi_s(u)$ are normalized, then $F_{m,s}(\cdot)$ and $F_{M,s}(\cdot)$ are also normalized, i.e., $F_{m,s}(1)=0$ and $F_{M,s}(1)=0$. Moreover, the functions f(u) and $\phi_s(u)$ are twice differentiable. Then in view of (2.4) and (3.1), we have

$$F''_{m,s}(u) = f''(u) - mu^{s-2} = u^{s-2} \left(u^{2-s} f''(u) - m \right) \ge 0$$

and

$$F_{M,s}''(u) = Mu^{s-2} - f''(u) = u^{s-2} \left(M - u^{2-s} f''(u) \right) \ge 0,$$

for all $u \in (r, R)$ and $s \in \mathbb{R}$. Thus the functions $F_{m,s}(\cdot)$ and $F_{M,s}(\cdot)$ are convex on (r, R).

We have seen above that the real mappings $F_{m,s}(\cdot)$ and $F_{M,s}(\cdot)$ defined over \mathbb{R}_+ given by (3.3) and (3.4) respectively are normalized, twice differentiable and convex on (r, R). Applying the r.h.s. of the inequality (2.6), we have

(3.5)
$$C_{F_{m,s}}(P||Q) \le \beta_{F_{m,s}}(r,R),$$

and

(3.6)
$$C_{F_{m,s}}(P||Q) \le \beta_{F_{M,s}}(r,R),$$



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respectively.

Moreover,

(3.7)
$$C_{F_{m,s}}(P||Q) = C_f(P||Q) - m\Phi_s(P||Q),$$

and

(3.8)
$$C_{F_{M_s}}(P||Q) = M\Phi_s(P||Q) - C_f(P||Q).$$

In view of (3.5) and (3.7), we have

$$C_f(P||Q) - m\Phi_s(P||Q) \le \beta_{F_{m,s}}(r,R),$$

i.e.,

$$C_f(P||Q) - m\Phi_s(P||Q) \le \beta_f(r,R) - m\phi_s(r,R)$$

i.e.,

$$m\left[\phi_s(r,R) - \Phi_s(P||Q)\right] \le \beta_f(r,R) - C_f(P||Q).$$

Thus, we have the l.h.s. of the inequality (3.2).

Again in view of (3.6) and (3.8), we have

$$M\Phi_s(P||Q) - C_f(P||Q) \le \beta_{F_{M,s}}(r,R),$$

i.e.,

$$M\Phi_s(P||Q) - C_f(P||Q) \le M\phi_s(r,R) - \beta_f(r,R),$$

i.e.,

$$\beta_f(r,R) - C_f(P||Q) \le M \left[\phi_s(r,R) - \Phi_s(P||Q)\right].$$

Thus we have the r.h.s. of the inequality (3.2).



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Remark 3.1. For similar kinds of results in comparing the f-divergence with Kullback-Leibler relative information see the work by Dragomir [10]. The case of Hellinger discrimination is discussed in Dragomir [6].

We shall now present some particular case of the Theorem 3.1.

3.1. Information Bounds in Terms of χ^2 -Divergence

In particular for s = 2, in Theorem 3.1, we have the following proposition:

Proposition 3.2. Let $f: I \subset \mathbb{R}_+ \to \mathbb{R}$ the generating mapping be normalized, i.e., f(1) = 0 and satisfy the assumptions:

- (i) f is twice differentiable on (r, R), where $0 < r \le 1 \le R < \infty$;
- (ii) there exists the real constants m, M with m < M such that

$$(3.9) m \le f''(x) \le M, \quad \forall x \in (r, R).$$

If $P, Q \in \Delta_n$ are discrete probability distributions satisfying the assumption

$$0 < r \le \frac{p_i}{q_i} \le R < \infty,$$

then we have the inequalities:

(3.10)
$$\frac{m}{2} \left[(R-1)(1-r) - \chi^2(P||Q) \right] \\ \leq \beta_f(r,R) - C_f(P||Q) \\ \leq \frac{M}{2} \left[(R-1)(1-r) - \chi^2(P||Q) \right],$$



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where $C_f(P||Q)$, $\beta_f(r,R)$ and $\chi^2(P||Q)$ are as given by (2.1), (2.8) and (1.6) respectively.

The above proposition was obtained by Dragomir in [9]. As a consequence of the above Proposition 3.2, we have the following result.

Result 3. Let $P, Q \in \Delta_n$ and $s \in \mathbb{R}$. Let there exist $r, R \ (0 < r \le 1 \le R < \infty)$ such that

$$0 < r \le \frac{p_i}{q_i} \le R < \infty, \quad \forall i \in \{1, 2, \dots, n\},$$

then in view of Proposition 3.2, we have

(3.11)
$$\frac{R^{s-2}}{2} \left[(R-1)(1-r) - \chi^2(P||Q) \right]$$

$$\leq \phi_s(r,R) - \Phi_s(P||Q)$$

$$\leq \frac{r^{s-2}}{2} \left[(R-1)(1-r) - \chi^2(P||Q) \right], \quad s \leq 2$$

and

(3.12)
$$\frac{r^{s-2}}{2} \left[(R-1)(1-r) - \chi^2(P||Q) \right] \\ \leq \phi_s(r,R) - \Phi_s(P||Q) \\ \leq \frac{R^{s-2}}{2} \left[(R-1)(1-r) - \chi^2(P||Q) \right], \quad s \geq 2.$$

Proof. Let us consider $f(u) = \phi_s(u)$, where $\phi_s(u)$ is as given by (2.2), then according to expression (2.4), we have

$$\phi_s''(u) = u^{s-2}.$$



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Now if $u \in [r, R] \subset (0, \infty)$, then we have

$$R^{s-2} \le \phi_s''(u) \le r^{s-2}, \quad s \le 2,$$

or accordingly, we have

(3.13)
$$\phi_s''(u) \begin{cases} \leq r^{s-2}, & s \leq 2; \\ \geq r^{s-2}, & s \geq 2 \end{cases}$$

and

(3.14)
$$\phi_s''(u) \begin{cases} \leq R^{s-2}, & s \geq 2; \\ \geq R^{s-2}, & s \leq 2, \end{cases}$$

where r and R are as defined above. Thus in view of (3.9), (3.13) and (3.14), we have the proof.

In view of Result 3, we have the following corollary.

Corollary 3.3. *Under the conditions of Result* 3, we have

(3.15)
$$\frac{1}{R^3} \left[(R-1)(1-r) - \chi^2(P||Q) \right] \\ \leq \frac{(R-1)(1-r)}{rR} - \chi^2(Q||P) \\ \leq \frac{1}{r^3} \left[(R-1)(1-r) - \chi^2(P||Q) \right],$$



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(3.16)
$$\frac{1}{2R^2} \left[(R-1)(1-r) - \chi^2(P||Q) \right] \\ \leq \frac{(R-1)\ln\frac{1}{r} + (1-r)\ln\frac{1}{R}}{R-r} - K(Q||P) \\ \leq \frac{1}{2r^2} \left[(R-1)(1-r) - \chi^2(P||Q) \right],$$

(3.17)
$$\frac{1}{2R} \left[(R-1)(1-r) - \chi^2(P||Q) \right] \\ \leq \frac{(R-1)r \ln r + (1-r)R \ln R}{R-r} - K(P||Q) \\ \leq \frac{1}{2r} \left[(R-1)(1-r) - \chi^2(P||Q) \right]$$

(3.18)
$$\frac{1}{8\sqrt{R^3}} \left[(R-1)(1-r) - \chi^2(P||Q) \right] \\ \leq \frac{\left(\sqrt{R}-1\right)(1-\sqrt{r})}{\sqrt{R}+\sqrt{r}} - h(P||Q) \\ \leq \frac{1}{8\sqrt{r^3}} \left[(R-1)(1-r) - \chi^2(P||Q) \right].$$

Proof. (3.15) follows by taking s=-1, (3.16) follows by taking s=0, (3.17) follows by taking s=1, (3.18) follows by taking $s=\frac{1}{2}$ in Result 3. While for s=2, we have equality sign.



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Proposition 3.4. Let $f: I \subset \mathbb{R}_+ \to \mathbb{R}$ the generating mapping be normalized, i.e., f(1) = 0 and satisfy the assumptions:

- (i) f is twice differentiable on (r, R), where $0 < r < 1 < R < \infty$;
- (ii) there exists the real constants m, M such that m < M and

$$(3.19) m \le x^3 f''(x) \le M, \quad \forall x \in (r, R).$$

If $P, Q \in \Delta_n$ are discrete probability distributions satisfying the assumption

$$0 < r \le \frac{p_i}{q_i} \le R < \infty,$$

then we have the inequalities:

(3.20)
$$\frac{m}{2} \left[\frac{(R-1)(1-r)}{rR} - \chi^{2}(Q||P) \right] \\ \leq \beta_{f}(r,R) - C_{f}(P||Q) \\ \leq \frac{m}{2} \left[\frac{(R-1)(1-r)}{rR} - \chi^{2}(Q||P) \right],$$

where $C_f(P||Q)$, $\beta_f(r,R)$ and $\chi^2(Q||P)$ are as given by (2.1), (2.8) and (1.7) respectively.

As a consequence of above proposition, we have the following result.

Result 4. Let $P, Q \in \Delta_n$ and $s \in \mathbb{R}$. Let there exist $r, R \ (0 < r \le 1 \le R < \infty)$ such that

$$0 < r \le \frac{p_i}{q_i} \le R < \infty, \quad \forall i \in \{1, 2, \dots, n\},$$



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then in view of Proposition 3.4, we have

(3.21)
$$\frac{R^{s+1}}{2} \left[\frac{(R-1)(1-r)}{rR} - \chi^2(Q||P) \right] \\ \leq \phi_s(r,R) - \Phi_s(P||Q) \\ \leq \frac{r^{s+1}}{2} \left[\frac{(R-1)(1-r)}{rR} - \chi^2(Q||P) \right], \quad s \leq -1$$

and

(3.22)
$$\frac{r^{s+1}}{2} \left[\frac{(R-1)(1-r)}{rR} - \chi^2(Q||P) \right] \\ \leq \phi_s(r,R) - \Phi_s(P||Q) \\ \leq \frac{R^{s+1}}{2} \left[\frac{(R-1)(1-r)}{rR} - \chi^2(Q||P) \right], \quad s \geq -1.$$

Proof. Let us consider $f(u) = \phi_s(u)$, where $\phi_s(u)$ is as given by (2.2), then according to expression (2.4), we have

$$\phi_s''(u) = u^{s-2}.$$

Let us define the function $g:[r,R]\to\mathbb{R}$ such that $g(u)=u^3\phi_s''(u)=u^{s+1}$, then we have

(3.23)
$$\sup_{u \in [r,R]} g(u) = \begin{cases} R^{s+1}, & s \ge -1; \\ r^{s+1}, & s \le -1 \end{cases}$$



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(3.24)
$$\inf_{u \in [r,R]} g(u) = \begin{cases} r^{s+1}, & s \ge -1; \\ R^{s+1}, & s \le -1. \end{cases}$$

In view of (3.23), (3.24) and Proposition 3.4, we have the proof of the result.

In view of Result 4, we have the following corollary.

Corollary 3.5. *Under the conditions of Result* 4, *we have*

(3.25)
$$\frac{r}{2} \left[\frac{(R-1)(1-r)}{rR} - \chi^{2}(Q||P) \right]$$

$$\leq \frac{(R-1)\ln\frac{1}{r} + (1-r)\ln\frac{1}{R}}{R-r} - K(Q||P)$$

$$\leq \frac{R}{2} \left[\frac{(R-1)(1-r)}{rR} - \chi^{2}(Q||P) \right],$$

(3.26)
$$\frac{r^2}{2} \left[\frac{(R-1)(1-r)}{rR} - \chi^2(Q||P) \right]$$

$$\leq \frac{(R-1)r \ln r + (1-r)R \ln R}{R-r} - K(P||Q)$$

$$\leq \frac{R^2}{2} \left[\frac{(R-1)(1-r)}{rR} - \chi^2(Q||P) \right],$$



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(3.27)
$$\frac{\sqrt{r^3}}{8} \left[\frac{(R-1)(1-r)}{rR} - \chi^2(Q||P) \right] \\ \leq \frac{\left(\sqrt{R}-1\right)(1-\sqrt{r})}{\sqrt{R}+\sqrt{r}} - h(P||Q) \\ \leq \frac{\sqrt{R^3}}{8} \left[\frac{(R-1)(1-r)}{rR} - \chi^2(Q||P) \right]$$

(3.28)
$$r^{3} \left[\frac{(R-1)(1-r)}{rR} - \chi^{2}(Q||P) \right]$$

$$\leq (R-1)(1-r) - \chi^{2}(P||Q)$$

$$\leq R^{3} \left[\frac{(R-1)(1-r)}{rR} - \chi^{2}(Q||P) \right].$$

Proof. (3.25) follows by taking s=0, (3.26) follows by taking s=1, (3.27) follows by taking $s=\frac{1}{2}$ and (3.28) follows by taking s=2 in Result 4. While for s=-1, we have equality sign.

3.2. Information Bounds in Terms of Kullback-Leibler Relative Information

In particular for s = 1, in the Theorem 3.1, we have the following proposition (see also Dragomir [10]).



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Proposition 3.6. Let $f: I \subset \mathbb{R}_+ \to \mathbb{R}$ the generating mapping be normalized, i.e., f(1) = 0 and satisfy the assumptions:

- (i) f is twice differentiable on (r, R), where $0 < r < 1 < R < \infty$;
- (ii) there exists the real constants m, M with m < M such that

$$(3.29) m \le x f''(x) \le M, \quad \forall x \in (r, R).$$

If $P, Q \in \Delta_n$ are discrete probability distributions satisfying the assumption

$$0 < r \le \frac{p_i}{q_i} \le R < \infty,$$

then we have the inequalities:

(3.30)
$$m \left[\frac{(R-1)r \ln r + (1-r)R \ln R}{R-r} - K(P||Q) \right]$$

$$\leq \beta_f(r,R) - C_f(P||Q)$$

$$\leq M \left[\frac{(R-1)r \ln r + (1-r)R \ln R}{R-r} - K(P||Q) \right],$$

where $C_f(P||Q)$, $\beta_f(r,R)$ and K(P||Q) as given by (2.1), (2.8) and (1.1) respectively.

In view of the above proposition, we have the following result.

Result 5. Let $P, Q \in \Delta_n$ and $s \in \mathbb{R}$. Let there exist $r, R \ (0 < r \le 1 \le R < \infty)$ such that

$$0 < r \le \frac{p_i}{q_i} \le R < \infty, \quad \forall i \in \{1, 2, \dots, n\},$$



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then in view of Proposition 3.6, we have

(3.31)
$$r^{s-1} \left[\frac{(R-1)r \ln r + (1-r)R \ln R}{R-r} - K(P||Q) \right]$$

$$\leq \phi_s(r,R) - \Phi_s(P||Q)$$

$$\leq R^{s-1} \left[\frac{(R-1)r \ln r + (1-r)R \ln R}{R-r} - K(P||Q) \right], \quad s \geq 1$$

and

(3.32)
$$R^{s-1} \left[\frac{(R-1)r \ln r + (1-r)R \ln R}{R-r} - K(P||Q) \right]$$

$$\leq \phi_s(r,R) - \Phi_s(P||Q)$$

$$\leq r^{s-1} \left[\frac{(R-1)r \ln r + (1-r)R \ln R}{R-r} - K(P||Q) \right], \quad s \leq 1.$$

Proof. Let us consider $f(u) = \phi_s(u)$, where $\phi_s(u)$ is as given by (2.2), then according to expression (2.4), we have

$$\phi_s''(u) = u^{s-2}.$$

Let us define the function $g:[r,R]\to\mathbb{R}$ such that $g(u)=\phi_s''(u)=u^{s-1}$, then we have

(3.33)
$$\sup_{u \in [r,R]} g(u) = \begin{cases} R^{s-1}, & s \ge 1; \\ r^{s-1}, & s \le 1 \end{cases}$$



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(3.34)
$$\inf_{u \in [r,R]} g(u) = \begin{cases} r^{s-1}, & s \ge 1; \\ R^{s-1}, & s \le 1. \end{cases}$$

In view of (3.33), (3.34) and Proposition 3.6 we have the proof of the result.

In view of Result 5, we have the following corollary.

Corollary 3.7. *Under the conditions of Result 5, we have*

(3.35)
$$\frac{2}{R^{2}} \left[\frac{(R-1)r \ln r + (1-r)R \ln R}{R-r} - K(P||Q) \right]$$

$$\leq \frac{(R-1)(1-r)}{rR} - \chi^{2}(Q||P)$$

$$\leq \frac{2}{r^{2}} \left[\frac{(R-1)r \ln r + (1-r)R \ln R}{R-r} - K(P||Q) \right],$$

(3.36)
$$\frac{1}{R} \left[\frac{(R-1)r \ln r + (1-r)R \ln R}{R-r} - K(P||Q) \right]$$

$$\leq \frac{(R-1) \ln \frac{1}{r} + (1-r) \ln \frac{1}{R}}{R-r} - K(Q||P)$$

$$\leq \frac{1}{r} \left[\frac{(R-1)r \ln r + (1-r)R \ln R}{R-r} - K(P||Q) \right],$$



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(3.37)
$$\frac{1}{4\sqrt{R}} \left[\frac{(R-1)r\ln r + (1-r)R\ln R}{R-r} - K(P||Q) \right] \\ \leq \frac{\left(\sqrt{R}-1\right)(1-\sqrt{r})}{\sqrt{R}+\sqrt{r}} - h(P||Q) \\ \leq \frac{1}{4\sqrt{r}} \left[\frac{(R-1)r\ln r + (1-r)R\ln R}{R-r} - K(P||Q) \right]$$

(3.38)
$$2r \left[\frac{(R-1)r \ln r + (1-r)R \ln R}{R-r} - K(P||Q) \right]$$

$$\leq (R-1)(1-r) - \chi^{2}(P||Q)$$

$$\leq 2R \left[\frac{(R-1)r \ln r + (1-r)R \ln R}{R-r} - K(P||Q) \right].$$

Proof. (3.35) follows by taking s = -1, (3.36) follows by taking s = 0, (3.37) follows by taking $s = \frac{1}{2}$ and (3.38) follows by taking s = 2 in Result 5. For s = 1, we have equality sign.

In particular, for s = 0 in Theorem 3.1, we have the following proposition:

Proposition 3.8. Let $f: I \subset \mathbb{R}_+ \to \mathbb{R}$ the generating mapping be normalized, i.e., f(1) = 0 and satisfy the assumptions:

(i) f is twice differentiable on (r, R), where $0 < r \le 1 \le R < \infty$;



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(ii) there exists the real constants m, M with m < M such that

$$(3.39) m \le x^2 f''(x) \le M, \quad \forall x \in (r, R).$$

If $P, Q \in \Delta_n$ are discrete probability distributions satisfying the assumption

$$0 < r \le \frac{p_i}{q_i} \le R < \infty,$$

then we have the inequalities:

(3.40)
$$m \left[\frac{(R-1)\ln\frac{1}{r} + (1-r)\ln\frac{1}{R}}{R-r} - K(Q||P) \right]$$

$$\leq \beta_f(r,R) - C_f(P||Q)$$

$$\leq M \left[\frac{(R-1)\ln\frac{1}{r} + (1-r)\ln\frac{1}{R}}{R-r} - K(Q||P) \right],$$

where $C_f(P||Q)$, $\beta_f(r,R)$ and K(Q||P) as given by (2.1), (2.8) and (1.1) respectively.

In view of Proposition 3.8, we have the following result.

Result 6. Let $P, Q \in \Delta_n$ and $s \in \mathbb{R}$. Let there exist $r, R \ (0 < r \le 1 \le R < \infty)$ such that

$$0 < r \le \frac{p_i}{q_i} \le R < \infty, \quad \forall i \in \{1, 2, \dots, n\},$$



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then in view of Proposition 3.8, we have

(3.41)
$$r^{s} \left[\frac{(R-1) \ln \frac{1}{r} + (1-r) \ln \frac{1}{R}}{R-r} - K(Q||P) \right]$$

$$\leq \phi_{s}(r,R) - \Phi_{s}(P||Q)$$

$$\leq R^{s} \left[\frac{(R-1) \ln \frac{1}{r} + (1-r) \ln \frac{1}{R}}{R-r} - K(Q||P) \right], \quad s \geq 0$$

and

(3.42)
$$R^{s} \left[\frac{(R-1)\ln\frac{1}{r} + (1-r)\ln\frac{1}{R}}{R-r} - K(Q||P) \right]$$

$$\leq \phi_{s}(r,R) - \Phi_{s}(P||Q)$$

$$\leq r^{s} \left[\frac{(R-1)\ln\frac{1}{r} + (1-r)\ln\frac{1}{R}}{R-r} - K(Q||P) \right], \quad s \leq 0.$$

Proof. Let us consider $f(u) = \phi_s(u)$, where $\phi_s(u)$ is as given by (2.2), then according to expression (2.4), we have

$$\phi_s''(u) = u^{s-2}.$$

Let us define the function $g:[r,R]\to\mathbb{R}$ such that $g(u)=u^2\phi_s''(u)=u^s$, then we have

(3.43)
$$\sup_{u \in [r,R]} g(u) = \begin{cases} R^s, & s \ge 0; \\ r^s, & s \le 0 \end{cases}$$



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(3.44)
$$\inf_{u \in [r,R]} g(u) = \begin{cases} r^s, & s \ge 0; \\ R^s, & s \le 0. \end{cases}$$

In view of (3.43), (3.44) and Proposition 3.8, we have the proof of the result.

In view of Result 6, we have the following corollary.

Corollary 3.9. *Under the conditions of Result* 6, we have

(3.45)
$$\frac{2}{R} \left[\frac{(R-1)\ln\frac{1}{r} + (1-r)\ln\frac{1}{R}}{R-r} - K(Q||P) \right]$$

$$\leq \frac{(R-1)(1-r)}{rR} - \chi^{2}(Q||P)$$

$$\leq \frac{2}{r} \left[\frac{(R-1)\ln\frac{1}{r} + (1-r)\ln\frac{1}{R}}{R-r} - K(Q||P) \right],$$

(3.46)
$$r \left[\frac{(R-1)\ln\frac{1}{r} + (1-r)\ln\frac{1}{R}}{R-r} - K(Q||P) \right]$$

$$\leq \frac{(R-1)r\ln r + (1-r)R\ln R}{R-r} - K(P||Q)$$

$$\leq R \left[\frac{(R-1)\ln\frac{1}{r} + (1-r)\ln\frac{1}{R}}{R-r} - K(Q||P) \right],$$



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(3.47)
$$\frac{\sqrt{r}}{4} \left[\frac{(R-1)\ln\frac{1}{r} + (1-r)\ln\frac{1}{R}}{R-r} - K(Q||P) \right]$$

$$\leq \frac{\left(\sqrt{R}-1\right)(1-\sqrt{r})}{\sqrt{R}+\sqrt{r}} - h(P||Q)$$

$$\leq \frac{\sqrt{R}}{4} \left[\frac{(R-1)\ln\frac{1}{r} + (1-r)\ln\frac{1}{R}}{R-r} - K(Q||P) \right]$$

(3.48)
$$2r^{2} \left[\frac{(R-1)\ln\frac{1}{r} + (1-r)\ln\frac{1}{R}}{R-r} - K(Q||P) \right]$$

$$\leq (R-1)(1-r) - \chi^{2}(P||Q)$$

$$\leq 2R^{2} \left[\frac{(R-1)\ln\frac{1}{r} + (1-r)\ln\frac{1}{R}}{R-r} - K(Q||P) \right].$$

Proof. (3.45) follows by taking s=-1, (3.46) follows by taking s=1, (3.47) follows by taking $s=\frac{1}{2}$ and (3.48) follows by taking s=2 in Result 6. For s=0, we have equality sign.

3.3. Information Bounds in Terms of Hellinger's Discrimination

In particular, for $s = \frac{1}{2}$ in Theorem 3.1, we have the following proposition (see also Dragomir [6]).



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Proposition 3.10. Let $f: I \subset \mathbb{R}_+ \to \mathbb{R}$ the generating mapping be normalized, i.e., f(1) = 0 and satisfy the assumptions:

- (i) f is twice differentiable on (r, R), where $0 < r \le 1 \le R < \infty$;
- (ii) there exists the real constants m, M with m < M such that

(3.49)
$$m \le x^{3/2} f''(x) \le M, \quad \forall x \in (r, R).$$

If $P, Q \in \Delta_n$ are discrete probability distributions satisfying the assumption

$$0 < r \le \frac{p_i}{q_i} \le R < \infty,$$

then we have the inequalities:

(3.50)
$$4m \left[\frac{\left(\sqrt{R} - 1\right)(1 - \sqrt{r})}{\sqrt{R} + \sqrt{r}} - h(P||Q) \right]$$
$$\leq \beta_f(r, R) - C_f(P||Q)$$
$$\leq 4M \left[\frac{\left(\sqrt{R} - 1\right)(1 - \sqrt{r})}{\sqrt{R} + \sqrt{r}} - h(P||Q) \right],$$

where $C_f(P||Q)$, $\beta_f(r,R)$ and h(P||Q) as given by (2.1), (2.8) and (1.5) respectively.

In view of Proposition 3.10, we have the following result.



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Result 7. Let $P, Q \in \Delta_n$ and $s \in \mathbb{R}$. Let there exist $r, R \ (0 < r \le 1 \le R < \infty)$ such that

$$0 < r \le \frac{p_i}{q_i} \le R < \infty, \quad \forall i \in \{1, 2, \dots, n\},$$

then in view of Proposition 3.10, we have

(3.51)
$$4r^{\frac{2s-1}{2}} \left[\frac{\left(\sqrt{R}-1\right)(1-\sqrt{r})}{\sqrt{R}+\sqrt{r}} - h(P||Q) \right] \\ \leq \phi_s(r,R) - \Phi_s(P||Q) \\ \leq 4R^{\frac{2s-1}{2}} \left[\frac{\left(\sqrt{R}-1\right)(1-\sqrt{r})}{\sqrt{R}+\sqrt{r}} - h(P||Q) \right], \quad s \geq \frac{1}{2}$$

and

(3.52)
$$4R^{\frac{2s-1}{2}} \left[\frac{\left(\sqrt{R}-1\right)(1-\sqrt{r})}{\sqrt{R}+\sqrt{r}} - h(P||Q) \right] \\ \leq \phi_s(r,R) - \Phi_s(P||Q) \\ \leq 4r^{\frac{2s-1}{2}} \left[\frac{\left(\sqrt{R}-1\right)(1-\sqrt{r})}{\sqrt{R}+\sqrt{r}} - h(P||Q) \right], \quad s \leq \frac{1}{2}.$$

Proof. Let the function $\phi_s(u)$ given by (3.29) be defined over [r, R]. Defining



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 $g(u) = u^{3/2} \phi_s''(u) = u^{\frac{2s-1}{2}}$, obviously we have

(3.53)
$$\sup_{u \in [r,R]} g(u) = \begin{cases} R^{\frac{2s-1}{2}}, & s \ge \frac{1}{2}; \\ r^{\frac{2s-1}{2}}, & s \le \frac{1}{2} \end{cases}$$

and

(3.54)
$$\inf_{u \in [r,R]} g(u) = \begin{cases} r^{\frac{2s-1}{2}}, & s \ge \frac{1}{2}; \\ R^{\frac{2s-1}{2}}, & s \le \frac{1}{2}. \end{cases}$$

In view of (3.53), (3.54) and Proposition 3.10, we get the proof of the result. \Box In view of Result 7, we have the following corollary.

Corollary 3.11. *Under the conditions of Result 7, we have*

(3.55)
$$\frac{8}{\sqrt{R^3}} \left[\frac{\left(\sqrt{R} - 1\right)(1 - \sqrt{r})}{\sqrt{R} + \sqrt{r}} - h(P||Q) \right]$$

$$\leq \frac{(R - 1)(1 - r)}{rR} - \chi^2(Q||P)$$

$$\leq \frac{8}{\sqrt{r^3}} \left[\frac{\left(\sqrt{R} - 1\right)(1 - \sqrt{r})}{\sqrt{R} + \sqrt{r}} - h(P||Q) \right],$$



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(3.56)
$$\frac{4}{\sqrt{R}} \left[\frac{\left(\sqrt{R} - 1\right)(1 - \sqrt{r})}{\sqrt{R} + \sqrt{r}} - h(P||Q) \right] \\ \leq \frac{(R - 1)\ln\frac{1}{r} + (1 - r)\ln\frac{1}{R}}{R - r} - K(Q||P) \\ \leq \frac{4}{\sqrt{r}} \left[\frac{\left(\sqrt{R} - 1\right)(1 - \sqrt{r})}{\sqrt{R} + \sqrt{r}} - h(P||Q) \right],$$

$$(3.57) 4\sqrt{r} \left[\frac{\left(\sqrt{R}-1\right)(1-\sqrt{r})}{\sqrt{R}+\sqrt{r}} - h(P||Q) \right]$$

$$\leq \frac{(R-1)r\ln r + (1-r)R\ln R}{R-r} - K(P||Q)$$

$$\leq 4\sqrt{R} \left[\frac{\left(\sqrt{R}-1\right)(1-\sqrt{r})}{\sqrt{R}+\sqrt{r}} - h(P||Q) \right]$$

(3.58)
$$8\sqrt{r^3} \left[\frac{\left(\sqrt{R} - 1\right)(1 - \sqrt{r})}{\sqrt{R} + \sqrt{r}} - h(P||Q) \right] \\ \leq (R - 1)(1 - r) - \chi^2(P||Q)$$



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$$\leq 8\sqrt{R^3} \left[\frac{\left(\sqrt{R} - 1\right)(1 - \sqrt{r})}{\sqrt{R} + \sqrt{r}} - h(P||Q) \right].$$

Proof. (3.55) follows by taking s = -1, (3.56) follows by taking s = 0, (3.57) follows by taking s = 1 and (3.58) follows by taking s = 2 in Result 7. For $s = \frac{1}{2}$, we have equality sign.



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