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## GENERALIZED RELATIVE INFORMATION AND INFORMATION INEQUALITIES

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Abstract

## 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 $\chi^{2}$-Divergence, Kullback-Leibler's relative information and Hellinger's discrimination.
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## 1. Introduction

Let

$$
\Delta_{n}=\left\{P=\left(p_{1}, p_{2}, \ldots, p_{n}\right) \mid p_{i}>0, \sum_{i=1}^{n} p_{i}=1\right\}, \quad n \geq 2
$$

be the set of complete finite discrete probability distributions.
The Kullback Leibler's [13] relative information is given by

$$
\begin{equation*}
K(P \| Q)=\sum_{i=1}^{n} p_{i} \ln \left(\frac{p_{i}}{q_{i}}\right), \tag{1.1}
\end{equation*}
$$

for all $P, Q \in \Delta_{n}$.
In $\Delta_{n}$, we have taken all $p_{i}>0$. If we take $p_{i} \geq 0, \forall i=1,2, \ldots, n$, 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

$$
\begin{equation*}
J(P \| Q)=K(P \| Q)+K(Q \| P)=\sum_{i=1}^{n}\left(p_{i}-q_{i}\right) \ln \left(\frac{p_{i}}{q_{i}}\right) \tag{1.2}
\end{equation*}
$$

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

$$
\begin{equation*}
K_{s}(P \| Q)=[s(s-1)]^{-1}\left[\sum_{i=1}^{n} p_{i}^{s} q_{i}^{1-s}-1\right], \quad s \neq 0,1 \tag{1.3}
\end{equation*}
$$

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In this case we have the following limiting cases

$$
\lim _{s \rightarrow 1} K_{s}(P \| Q)=K(P \| Q)
$$

and

$$
\lim _{s \rightarrow 0} K_{s}(P \| Q)=K(Q \| P)
$$

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

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

where

$$
\begin{equation*}
B(P \| Q)=\sum_{i=1}^{n} \sqrt{p_{i} q_{i}} \tag{1.4}
\end{equation*}
$$

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

$$
\begin{equation*}
h(P \| Q)=\frac{1}{2} \sum_{i=1}^{n}\left(\sqrt{p_{i}}-\sqrt{q_{i}}\right)^{2} \tag{1.5}
\end{equation*}
$$

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is famous as Hellinger's [11] discrimination.
(ii) When $s=2$, we have

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

where

$$
\begin{equation*}
\chi^{2}(P \| Q)=\sum_{i=1}^{n} \frac{\left(p_{i}-q_{i}\right)^{2}}{q_{i}}=\sum_{i=1}^{n} \frac{p_{i}^{2}}{q_{i}}-1, \tag{1.6}
\end{equation*}
$$

is the $\chi^{2}$-divergence (Pearson [16]).
(iii) When $s=-1$, we have

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

where

$$
\begin{equation*}
\chi^{2}(Q \| P)=\sum_{i=1}^{n} \frac{\left(p_{i}-q_{i}\right)^{2}}{p_{i}}=\sum_{i=1}^{n} \frac{q_{i}^{2}}{p_{i}}-1 \tag{1.7}
\end{equation*}
$$

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

$$
\Phi_{s}(P \| Q)= \begin{cases}K_{s}(P \| Q), & s \neq 0,1  \tag{1.8}\\ K(Q \| P), & s=0 \\ K(P \| Q), & s=1\end{cases}
$$

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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) $\Phi_{1 / 2}(P \| Q)=4[1-B(P \| Q)]=4 h(P \| Q)$.
(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 $f:[0, \infty) \rightarrow \mathbb{R}$, the $f$-divergence measure introduced by Csiszár [4] is given by

$$
\begin{equation*}
C_{f}(p, q)=\sum_{i=1}^{n} q_{i} f\left(\frac{p_{i}}{q_{i}}\right) \tag{2.1}
\end{equation*}
$$

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 $f:[0, \infty) \rightarrow \mathbb{R}$ 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 $f:[0, \infty) \rightarrow \mathbb{R}$ 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

$$
C_{f}(p, q) \geq Q_{n} f\left(\frac{P_{n}}{Q_{n}}\right)
$$

The equality sign holds for strictly convex functions iff

$$
\frac{p_{1}}{q_{i}}=\frac{p_{2}}{q_{2}}=\cdots=\frac{p_{n}}{q_{n}} .
$$

In particular, for all $P, Q \in \Delta_{n}$, we have

$$
C_{f}(P \| Q) \geq f(1)
$$

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

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

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

$$
C_{f}(P \| Q)=\Phi_{s}(P \| Q)= \begin{cases}K_{s}(P \| Q), & s \neq 0,1 \\ K(Q \| P), & s=0 \\ K(P \| Q), & s=1\end{cases}
$$

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Moreover,

$$
\phi_{s}^{\prime}(u)= \begin{cases}(s-1)^{-1}\left(u^{s-1}-1\right), & s \neq 0,1  \tag{2.3}\\ 1-u^{-1}, & s=0 \\ \ln u, & s=1\end{cases}
$$

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and

$$
\phi_{s}^{\prime \prime}(u)= \begin{cases}u^{s-2}, & s \neq 0,1  \tag{2.4}\\ u^{-2}, & s=0 \\ u^{-1}, & s=1\end{cases}
$$

Thus we have $\phi_{s}^{\prime \prime}(u)>0$ 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 $f: \mathbb{R}_{+} \rightarrow \mathbb{R}$ be differentiable convex and normalized i.e., $f(1)=0$. If $P, Q \in \Delta_{n}$ are such that

$$
0<r \leq \frac{p_{i}}{q_{i}} \leq R<\infty, \quad \forall i \in\{1,2, \ldots, n\}
$$

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

$$
\begin{equation*}
0 \leq C_{f}(P \| Q) \leq \frac{1}{4}(R-r)\left(f^{\prime}(R)-f^{\prime}(r)\right) \tag{2.5}
\end{equation*}
$$

$$
\begin{equation*}
0 \leq C_{f}(P \| Q) \leq \beta_{f}(r, R) \tag{2.6}
\end{equation*}
$$

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and

$$
\begin{align*}
0 & \leq \beta_{f}(r, R)-C_{f}(P \| Q)  \tag{2.7}\\
& \leq \frac{f^{\prime}(R)-f^{\prime}(r)}{R-r}\left[(R-1)(1-r)-\chi^{2}(P \| Q)\right] \\
& \leq \frac{1}{4}(R-r)\left(f^{\prime}(R)-f^{\prime}(r)\right)
\end{align*}
$$

where

$$
\begin{equation*}
\beta_{f}(r, R)=\frac{(R-1) f(r)+(1-r) f(R)}{R-r} \tag{2.8}
\end{equation*}
$$

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

$$
0<r \leq \frac{p_{i}}{q_{i}} \leq R<\infty, \quad \forall i \in\{1,2, \ldots, n\}
$$

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

$$
\begin{equation*}
0 \leq \Phi_{s}(P \| Q) \leq \mu_{s}(r, R) \tag{2.9}
\end{equation*}
$$

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$$
\begin{equation*}
0 \leq \Phi_{s}(P \| Q) \leq \phi_{s}(r, R) \tag{2.10}
\end{equation*}
$$

and

$$
\begin{align*}
0 & \leq \phi_{s}(r, R)-\Phi_{s}(P \| Q)  \tag{2.11}\\
& \leq k_{s}(r, R)\left[(R-1)(1-r)-\chi^{2}(P \| Q)\right] \\
& \leq \mu_{s}(r, R)
\end{align*}
$$

where

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

$$
\begin{align*}
\phi_{s}(r, R) & =\frac{(R-1) \phi_{s}(r)+(1-r) \phi_{s}(R)}{R-r}  \tag{2.13}\\
& = \begin{cases}\frac{(R-1)\left(r^{s}-1\right)+(1-r)\left(R^{s}-1\right)}{(R-r) s(s-1)}, & s \neq 0,1 \\
\frac{(R-1) \ln \frac{1}{r}+(1-r) \ln \frac{1}{R}}{(R-r)}, & s=0 \\
\frac{(R-1) r \ln r+(1-r) R \ln R}{(R-r)}, & s=1\end{cases}
\end{align*}
$$

and

$$
k_{s}(r, R)=\frac{\phi_{s}^{\prime}(R)-\phi_{s}^{\prime}(r)}{R-r}= \begin{cases}\frac{R^{s-1}-r^{s-1}}{(R-r)(s-1)}, & s \neq 1  \tag{2.14}\\ \frac{\ln R-\ln r}{R-r}, & s=1\end{cases}
$$

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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

$$
\begin{align*}
& 0 \leq \chi^{2}(Q \| P) \leq \frac{1}{4}(R+r)\left(\frac{R-r}{r R}\right)^{2}  \tag{2.15}\\
& 0 \leq K(Q \| P) \leq \frac{(R-r)^{2}}{4 R r} \tag{2.16}
\end{align*}
$$

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$$
\begin{align*}
& 0 \leq K(P \| Q) \leq \frac{1}{4}(R-r) \ln \left(\frac{R}{r}\right)  \tag{2.17}\\
& 0 \leq h(P \| Q) \leq \frac{(R-r)(\sqrt{R}-\sqrt{r})}{8 \sqrt{R r}} \tag{2.18}
\end{align*}
$$

and

$$
\begin{equation*}
0 \leq \chi^{2}(P \| Q) \leq \frac{1}{2}(R-r)^{2} \tag{2.19}
\end{equation*}
$$

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

$$
\begin{align*}
& 0 \leq \chi^{2}(Q \| P) \leq \frac{(R-1)(1-r)}{r R}  \tag{2.20}\\
& 0 \leq K(Q \| P) \leq \frac{(R-1) \ln \frac{1}{r}+(1-r) \ln \frac{1}{R}}{R-r}  \tag{2.21}\\
& 0 \leq K(P \| Q) \leq \frac{(R-1) r \ln r+(1-r) R \ln R}{R-r} \tag{2.22}
\end{align*}
$$

$$
\begin{equation*}
0 \leq h(P \| Q) \leq \frac{(\sqrt{R}-1)(1-\sqrt{r})}{\sqrt{R}+\sqrt{r}} \tag{2.23}
\end{equation*}
$$

and

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$$
\begin{equation*}
0 \leq \chi^{2}(P \| Q) \leq(R-1)(1-r) \tag{2.24}
\end{equation*}
$$

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:

$$
\begin{equation*}
0 \leq J(P \| Q) \leq \min \left\{t_{1}(r, R), t_{2}(r, R)\right\} \tag{2.25}
\end{equation*}
$$

where

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

and

$$
t_{2}(r, R)=(R-1)(1-r)\left(L_{-1}(r, R)\right)^{-1}
$$

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

$$
\begin{align*}
0 & \leq \frac{(R-1)(1-r)}{r R}-\chi^{2}(Q \| P)  \tag{2.26}\\
& \leq \frac{R+r}{(r R)^{2}}\left[(R-1)(1-r)-\chi^{2}(P \| Q)\right] \\
0 & \leq \frac{(R-1) \ln \frac{1}{r}+(1-r) \ln \frac{1}{R}}{R-r}-K(Q \| P)  \tag{2.27}\\
& \leq \frac{1}{r R}\left[(R-1)(1-r)-\chi^{2}(P \| Q)\right]
\end{align*}
$$

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$$
\begin{align*}
0 & \leq \frac{(R-1) r \ln r+(1-r) R \ln R}{R-r}-K(P \| Q)  \tag{2.28}\\
& \leq \frac{\ln R-\ln r}{R-r}\left[(R-1)(1-r)-\chi^{2}(P \| Q)\right]
\end{align*}
$$

and
(2.29) $0 \leq \frac{(\sqrt{R}-1)(1-\sqrt{r})}{(\sqrt{R}+\sqrt{r})}-h(P \| Q)$

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

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 $\chi^{2}-$ 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 $\chi^{2}$-divergence, Kullback-Leibler's relative information and Hellinger's discrimination.

Theorem 3.1. Let $f: I \subset \mathbb{R}_{+} \rightarrow \mathbb{R}$ the generating mapping be normalized, i.e., $f(1)=0$ and satisfy the assumptions:
(i) $f$ is twice differentiable on $(r, R)$, where $0 \leq r \leq 1 \leq R \leq \infty$;
(ii) there exists the real constants $m, M$ with $m<M$ such that

$$
\begin{equation*}
m \leq x^{2-s} f^{\prime \prime}(x) \leq M, \quad \forall x \in(r, R), \quad s \in \mathbb{R} \tag{3.1}
\end{equation*}
$$

If $P, Q \in \Delta_{n}$ are discrete probability distributions satisfying the assumption

$$
0<r \leq \frac{p_{i}}{q_{i}} \leq R<\infty
$$

then we have the inequalities:

$$
\begin{align*}
m\left[\phi_{s}(r, R)-\Phi_{s}(P \| Q)\right] & \leq \beta_{f}(r, R)-C_{f}(P \| Q)  \tag{3.2}\\
& \leq M\left[\phi_{s}(r, R)-\Phi_{s}(P \| Q)\right]
\end{align*}
$$

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

$$
\begin{equation*}
F_{m, s}(u)=f(u)-m \phi_{s}(u), \tag{3.3}
\end{equation*}
$$

and

$$
\begin{equation*}
F_{M, s}(u)=M \phi_{s}(u)-f(u) \tag{3.4}
\end{equation*}
$$

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

$$
F_{m, s}^{\prime \prime}(u)=f^{\prime \prime}(u)-m u^{s-2}=u^{s-2}\left(u^{2-s} f^{\prime \prime}(u)-m\right) \geq 0
$$

and

$$
F_{M, s}^{\prime \prime}(u)=M u^{s-2}-f^{\prime \prime}(u)=u^{s-2}\left(M-u^{2-s} f^{\prime \prime}(u)\right) \geq 0
$$

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

$$
\begin{equation*}
C_{F_{m, s}}(P \| Q) \leq \beta_{F_{m, s}}(r, R) \tag{3.5}
\end{equation*}
$$

and

$$
\begin{equation*}
C_{F_{m, s}}(P \| Q) \leq \beta_{F_{M, s}}(r, R) \tag{3.6}
\end{equation*}
$$

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respectively.
Moreover,

$$
\begin{equation*}
C_{F_{m, s}}(P \| Q)=C_{f}(P \| Q)-m \Phi_{s}(P \| Q), \tag{3.7}
\end{equation*}
$$

and

$$
\begin{equation*}
C_{F_{M, s}}(P \| Q)=M \Phi_{s}(P \| Q)-C_{f}(P \| Q) \tag{3.8}
\end{equation*}
$$

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

$$
C_{f}(P \| Q)-m \Phi_{s}(P \| Q) \leq \beta_{F_{m, s}}(r, R)
$$

i.e.,

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

i.e.,

$$
m\left[\phi_{s}(r, R)-\Phi_{s}(P \| Q)\right] \leq \beta_{f}(r, R)-C_{f}(P \| Q) .
$$

Thus, we have the l.h.s. of the inequality (3.2).
Again in view of (3.6) and (3.8), we have

$$
M \Phi_{s}(P \| Q)-C_{f}(P \| Q) \leq \beta_{F_{M, s}}(r, R),
$$

i.e.,

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

i.e.,

$$
\beta_{f}(r, R)-C_{f}(P \| Q) \leq M\left[\phi_{s}(r, R)-\Phi_{s}(P \| Q)\right]
$$

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

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 $\chi^{2}$-Divergence

In particular for $s=2$, in Theorem 3.1, we have the following proposition:
Proposition 3.2. Let $f: I \subset \mathbb{R}_{+} \rightarrow \mathbb{R}$ the generating mapping be normalized, i.e., $f(1)=0$ and satisfy the assumptions:
(i) $f$ is twice differentiable on $(r, R)$, where $0<r \leq 1 \leq R<\infty$;
(ii) there exists the real constants $m, M$ with $m<M$ such that

$$
\begin{equation*}
m \leq f^{\prime \prime}(x) \leq M, \quad \forall x \in(r, R) \tag{3.9}
\end{equation*}
$$

If $P, Q \in \Delta_{n}$ are discrete probability distributions satisfying the assumption

$$
0<r \leq \frac{p_{i}}{q_{i}} \leq R<\infty
$$

then we have the inequalities:

$$
\begin{align*}
\frac{m}{2} & {\left[(R-1)(1-r)-\chi^{2}(P \| Q)\right] }  \tag{3.10}\\
& \leq \beta_{f}(r, R)-C_{f}(P \| Q) \\
& \leq \frac{M}{2}\left[(R-1)(1-r)-\chi^{2}(P \| Q)\right]
\end{align*}
$$

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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 $r, R(0<r \leq 1 \leq R<\infty)$ such that

$$
0<r \leq \frac{p_{i}}{q_{i}} \leq R<\infty, \quad \forall i \in\{1,2, \ldots, n\}
$$

then in view of Proposition 3.2, we have

$$
\begin{align*}
\frac{R^{s-2}}{2} & {\left[(R-1)(1-r)-\chi^{2}(P \| Q)\right] }  \tag{3.11}\\
& \leq \phi_{s}(r, R)-\Phi_{s}(P \| Q) \\
& \leq \frac{r^{s-2}}{2}\left[(R-1)(1-r)-\chi^{2}(P \| Q)\right], \quad s \leq 2
\end{align*}
$$

and

$$
\begin{align*}
\frac{r^{s-2}}{2} & {\left[(R-1)(1-r)-\chi^{2}(P \| Q)\right] }  \tag{3.12}\\
& \leq \phi_{s}(r, R)-\Phi_{s}(P \| Q) \\
& \leq \frac{R^{s-2}}{2}\left[(R-1)(1-r)-\chi^{2}(P \| Q)\right], \quad s \geq 2
\end{align*}
$$

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

$$
\phi_{s}^{\prime \prime}(u)=u^{s-2}
$$

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

$$
R^{s-2} \leq \phi_{s}^{\prime \prime}(u) \leq r^{s-2}, \quad s \leq 2
$$

or accordingly, we have

$$
\phi_{s}^{\prime \prime}(u) \begin{cases}\leq r^{s-2}, & s \leq 2  \tag{3.13}\\ \geq r^{s-2}, & s \geq 2\end{cases}
$$

and

$$
\phi_{s}^{\prime \prime}(u) \begin{cases}\leq R^{s-2}, & s \geq 2  \tag{3.14}\\ \geq R^{s-2}, & s \leq 2\end{cases}
$$

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

$$
\begin{align*}
\frac{1}{R^{3}} & {\left[(R-1)(1-r)-\chi^{2}(P \| Q)\right] }  \tag{3.15}\\
& \leq \frac{(R-1)(1-r)}{r R}-\chi^{2}(Q \| P) \\
& \leq \frac{1}{r^{3}}\left[(R-1)(1-r)-\chi^{2}(P \| Q)\right]
\end{align*}
$$

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

$$
\begin{align*}
& \frac{1}{2 R}\left[(R-1)(1-r)-\chi^{2}(P \| Q)\right]  \tag{3.17}\\
& \quad \leq \frac{(R-1) r \ln r+(1-r) R \ln R}{R-r}-K(P \| Q) \\
& \quad \leq \frac{1}{2 r}\left[(R-1)(1-r)-\chi^{2}(P \| Q)\right]
\end{align*}
$$

and

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

Proof. (3.15) follows by taking $s=-1$, (3.16) follows by taking $s=0$, (3.17)

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Page 22 of 42 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.

Proposition 3.4. Let $f: I \subset \mathbb{R}_{+} \rightarrow \mathbb{R}$ the generating mapping be normalized, i.e., $f(1)=0$ and satisfy the assumptions:
(i) $f$ is twice differentiable on $(r, R)$, where $0<r \leq 1 \leq R<\infty$;
(ii) there exists the real constants $m$, Msuch that $m<M$ and

$$
\begin{equation*}
m \leq x^{3} f^{\prime \prime}(x) \leq M, \quad \forall x \in(r, R) \tag{3.19}
\end{equation*}
$$

If $P, Q \in \Delta_{n}$ are discrete probability distributions satisfying the assumption

$$
0<r \leq \frac{p_{i}}{q_{i}} \leq R<\infty
$$

then we have the inequalities:

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

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 $r, R(0<r \leq 1 \leq R<\infty)$ such that

$$
0<r \leq \frac{p_{i}}{q_{i}} \leq R<\infty, \quad \forall i \in\{1,2, \ldots, n\}
$$

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then in view of Proposition 3.4, we have

$$
\begin{align*}
\frac{R^{s+1}}{2} & {\left[\frac{(R-1)(1-r)}{r R}-\chi^{2}(Q \| P)\right] }  \tag{3.21}\\
& \leq \phi_{s}(r, R)-\Phi_{s}(P \| Q) \\
& \leq \frac{r^{s+1}}{2}\left[\frac{(R-1)(1-r)}{r R}-\chi^{2}(Q \| P)\right], \quad s \leq-1
\end{align*}
$$

and

$$
\begin{align*}
\frac{r^{s+1}}{2} & {\left[\frac{(R-1)(1-r)}{r R}-\chi^{2}(Q \| P)\right] }  \tag{3.22}\\
& \leq \phi_{s}(r, R)-\Phi_{s}(P \| Q) \\
& \leq \frac{R^{s+1}}{2}\left[\frac{(R-1)(1-r)}{r R}-\chi^{2}(Q \| P)\right], s \geq-1
\end{align*}
$$

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

$$
\phi_{s}^{\prime \prime}(u)=u^{s-2} .
$$

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

$$
\sup _{u \in[r, R]} g(u)= \begin{cases}R^{s+1}, & s \geq-1  \tag{3.23}\\ r^{s+1}, & s \leq-1\end{cases}
$$

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and

$$
\inf _{u \in[r, R]} g(u)= \begin{cases}r^{s+1}, & s \geq-1  \tag{3.24}\\ R^{s+1}, & s \leq-1\end{cases}
$$

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

$$
\begin{align*}
\frac{r}{2} & {\left[\frac{(R-1)(1-r)}{r R}-\chi^{2}(Q \| P)\right] }  \tag{3.25}\\
& \leq \frac{(R-1) \ln \frac{1}{r}+(1-r) \ln \frac{1}{R}}{R-r}-K(Q \| P) \\
& \leq \frac{R}{2}\left[\frac{(R-1)(1-r)}{r R}-\chi^{2}(Q \| P)\right]
\end{align*}
$$

$$
\begin{align*}
\frac{r^{2}}{2} & {\left[\frac{(R-1)(1-r)}{r R}-\chi^{2}(Q \| P)\right] }  \tag{3.26}\\
& \leq \frac{(R-1) r \ln r+(1-r) R \ln R}{R-r}-K(P \| Q) \\
& \leq \frac{R^{2}}{2}\left[\frac{(R-1)(1-r)}{r R}-\chi^{2}(Q \| P)\right]
\end{align*}
$$

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$$
\begin{align*}
& \frac{\sqrt{r^{3}}}{8}\left[\frac{(R-1)(1-r)}{r R}-\chi^{2}(Q \| P)\right]  \tag{3.27}\\
& \quad \leq \frac{(\sqrt{R}-1)(1-\sqrt{r})}{\sqrt{R}+\sqrt{r}}-h(P \| Q) \\
& \quad \leq \frac{\sqrt{R^{3}}}{8}\left[\frac{(R-1)(1-r)}{r R}-\chi^{2}(Q \| P)\right]
\end{align*}
$$

and

$$
\begin{align*}
r^{3} & {\left[\frac{(R-1)(1-r)}{r R}-\chi^{2}(Q \| P)\right] }  \tag{3.28}\\
& \leq(R-1)(1-r)-\chi^{2}(P \| Q) \\
& \leq R^{3}\left[\frac{(R-1)(1-r)}{r R}-\chi^{2}(Q \| P)\right] .
\end{align*}
$$

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 $f: I \subset \mathbb{R}_{+} \rightarrow \mathbb{R}$ the generating mapping be normalized, i.e., $f(1)=0$ and satisfy the assumptions:
(i) $f$ is twice differentiable on $(r, R)$, where $0<r \leq 1 \leq R<\infty$;
(ii) there exists the real constants $m, M$ with $m<M$ such that

$$
\begin{equation*}
m \leq x f^{\prime \prime}(x) \leq M, \quad \forall x \in(r, R) \tag{3.29}
\end{equation*}
$$

If $P, Q \in \Delta_{n}$ are discrete probability distributions satisfying the assumption

$$
0<r \leq \frac{p_{i}}{q_{i}} \leq R<\infty
$$

then we have the inequalities:

$$
\begin{align*}
& m\left[\frac{(R-1) r \ln r+(1-r) R \ln R}{R-r}-K(P \| Q)\right]  \tag{3.30}\\
& \quad \leq \beta_{f}(r, R)-C_{f}(P \| Q) \\
& \quad \leq M\left[\frac{(R-1) r \ln r+(1-r) R \ln R}{R-r}-K(P \| Q)\right]
\end{align*}
$$

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 $r, R(0<r \leq 1 \leq R<\infty)$ such that

$$
0<r \leq \frac{p_{i}}{q_{i}} \leq R<\infty, \quad \forall i \in\{1,2, \ldots, n\}
$$

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then in view of Proposition 3.6, we have
(3.31) $\quad r^{s-1}\left[\frac{(R-1) r \ln r+(1-r) R \ln R}{R-r}-K(P \| Q)\right]$

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

and

$$
\begin{align*}
R^{s-1} & {\left[\frac{(R-1) r \ln r+(1-r) R \ln R}{R-r}-K(P \| Q)\right] }  \tag{3.32}\\
& \leq \phi_{s}(r, R)-\Phi_{s}(P \| Q) \\
& \leq r^{s-1}\left[\frac{(R-1) r \ln r+(1-r) R \ln R}{R-r}-K(P \| Q)\right], \quad s \leq 1
\end{align*}
$$

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

$$
\phi_{s}^{\prime \prime}(u)=u^{s-2}
$$

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

$$
\sup _{u \in[r, R]} g(u)= \begin{cases}R^{s-1}, & s \geq 1  \tag{3.33}\\ r^{s-1}, & s \leq 1\end{cases}
$$

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and

$$
\inf _{u \in[r, R]} g(u)= \begin{cases}r^{s-1}, & s \geq 1  \tag{3.34}\\ R^{s-1}, & s \leq 1\end{cases}
$$

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

$$
\begin{align*}
\frac{2}{R^{2}} & {\left[\frac{(R-1) r \ln r+(1-r) R \ln R}{R-r}-K(P \| Q)\right] }  \tag{3.35}\\
& \leq \frac{(R-1)(1-r)}{r R}-\chi^{2}(Q \| P) \\
& \leq \frac{2}{r^{2}}\left[\frac{(R-1) r \ln r+(1-r) R \ln R}{R-r}-K(P \| Q)\right]
\end{align*}
$$

$$
\begin{align*}
& \frac{1}{R}\left[\frac{(R-1) r \ln r+(1-r) R \ln R}{R-r}-K(P \| Q)\right]  \tag{3.36}\\
& \quad \leq \frac{(R-1) \ln \frac{1}{r}+(1-r) \ln \frac{1}{R}}{R-r}-K(Q \| P) \\
& \quad \leq \frac{1}{r}\left[\frac{(R-1) r \ln r+(1-r) R \ln R}{R-r}-K(P \| Q)\right]
\end{align*}
$$

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

and

$$
\begin{align*}
2 r & {\left[\frac{(R-1) r \ln r+(1-r) R \ln R}{R-r}-K(P \| Q)\right] }  \tag{3.38}\\
& \leq(R-1)(1-r)-\chi^{2}(P \| Q) \\
& \leq 2 R\left[\frac{(R-1) r \ln r+(1-r) R \ln R}{R-r}-K(P \| Q)\right]
\end{align*}
$$

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 $f: I \subset \mathbb{R}_{+} \rightarrow \mathbb{R}$ the generating mapping be normalized, i.e., $f(1)=0$ and satisfy the assumptions:
(i) $f$ is twice differentiable on $(r, R)$, where $0<r \leq 1 \leq R<\infty$;

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

$$
\begin{equation*}
m \leq x^{2} f^{\prime \prime}(x) \leq M, \quad \forall x \in(r, R) \tag{3.39}
\end{equation*}
$$

If $P, Q \in \Delta_{n}$ are discrete probability distributions satisfying the assumption

$$
0<r \leq \frac{p_{i}}{q_{i}} \leq R<\infty
$$

then we have the inequalities:

$$
\begin{align*}
& m\left[\frac{(R-1) \ln \frac{1}{r}+(1-r) \ln \frac{1}{R}}{R-r}-K(Q \| P)\right]  \tag{3.40}\\
& \quad \leq \beta_{f}(r, R)-C_{f}(P \| Q) \\
& \quad \leq M\left[\frac{(R-1) \ln \frac{1}{r}+(1-r) \ln \frac{1}{R}}{R-r}-K(Q \| P)\right]
\end{align*}
$$

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 $r, R(0<r \leq 1 \leq R<\infty)$ such that

$$
0<r \leq \frac{p_{i}}{q_{i}} \leq R<\infty, \quad \forall i \in\{1,2, \ldots, n\}
$$

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then in view of Proposition 3.8, we have

$$
\begin{align*}
r^{s} & {\left[\frac{(R-1) \ln \frac{1}{r}+(1-r) \ln \frac{1}{R}}{R-r}-K(Q \| P)\right] }  \tag{3.41}\\
& \leq \phi_{s}(r, R)-\Phi_{s}(P \| Q) \\
& \leq R^{s}\left[\frac{(R-1) \ln \frac{1}{r}+(1-r) \ln \frac{1}{R}}{R-r}-K(Q \| P)\right], s \geq 0
\end{align*}
$$

and

$$
\begin{align*}
R^{s} & {\left[\frac{(R-1) \ln \frac{1}{r}+(1-r) \ln \frac{1}{R}}{R-r}-K(Q \| P)\right] }  \tag{3.42}\\
& \leq \phi_{s}(r, R)-\Phi_{s}(P \| Q) \\
& \leq r^{s}\left[\frac{(R-1) \ln \frac{1}{r}+(1-r) \ln \frac{1}{R}}{R-r}-K(Q \| P)\right], \quad s \leq 0
\end{align*}
$$

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

$$
\phi_{s}^{\prime \prime}(u)=u^{s-2}
$$

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

$$
\sup _{u \in[r, R]} g(u)= \begin{cases}R^{s}, & s \geq 0  \tag{3.43}\\ r^{s}, & s \leq 0\end{cases}
$$

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and

$$
\inf _{u \in[r, R]} g(u)= \begin{cases}r^{s}, & s \geq 0  \tag{3.44}\\ R^{s}, & s \leq 0\end{cases}
$$

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

$$
\begin{align*}
\frac{2}{R} & {\left[\frac{(R-1) \ln \frac{1}{r}+(1-r) \ln \frac{1}{R}}{R-r}-K(Q \| P)\right] }  \tag{3.45}\\
& \leq \frac{(R-1)(1-r)}{r R}-\chi^{2}(Q \| P) \\
& \leq \frac{2}{r}\left[\frac{(R-1) \ln \frac{1}{r}+(1-r) \ln \frac{1}{R}}{R-r}-K(Q \| P)\right]
\end{align*}
$$

$$
\begin{align*}
& r\left[\frac{(R-1) \ln \frac{1}{r}+(1-r) \ln \frac{1}{R}}{R-r}-K(Q \| P)\right]  \tag{3.46}\\
& \quad \leq \frac{(R-1) r \ln r+(1-r) R \ln R}{R-r}-K(P \| Q) \\
& \quad \leq R\left[\frac{(R-1) \ln \frac{1}{r}+(1-r) \ln \frac{1}{R}}{R-r}-K(Q \| P)\right]
\end{align*}
$$

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$$
\begin{align*}
& \frac{\sqrt{r}}{4}\left[\frac{(R-1) \ln \frac{1}{r}+(1-r) \ln \frac{1}{R}}{R-r}-K(Q \| P)\right]  \tag{3.47}\\
& \quad \leq \frac{(\sqrt{R}-1)(1-\sqrt{r})}{\sqrt{R}+\sqrt{r}}-h(P \| Q) \\
& \quad \leq \frac{\sqrt{R}}{4}\left[\frac{(R-1) \ln \frac{1}{r}+(1-r) \ln \frac{1}{R}}{R-r}-K(Q \| P)\right]
\end{align*}
$$

and

$$
\begin{align*}
2 r^{2} & {\left[\frac{(R-1) \ln \frac{1}{r}+(1-r) \ln \frac{1}{R}}{R-r}-K(Q \| P)\right] }  \tag{3.48}\\
& \leq(R-1)(1-r)-\chi^{2}(P \| Q) \\
& \leq 2 R^{2}\left[\frac{(R-1) \ln \frac{1}{r}+(1-r) \ln \frac{1}{R}}{R-r}-K(Q \| P)\right] .
\end{align*}
$$

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 $f: I \subset \mathbb{R}_{+} \rightarrow \mathbb{R}$ the generating mapping be normalized, i.e., $f(1)=0$ and satisfy the assumptions:
(i) $f$ is twice differentiable on $(r, R)$, where $0<r \leq 1 \leq R<\infty$;
(ii) there exists the real constants $m, M$ with $m<M$ such that

$$
\begin{equation*}
m \leq x^{3 / 2} f^{\prime \prime}(x) \leq M, \quad \forall x \in(r, R) \tag{3.49}
\end{equation*}
$$

If $P, Q \in \Delta_{n}$ are discrete probability distributions satisfying the assumption

$$
0<r \leq \frac{p_{i}}{q_{i}} \leq R<\infty,
$$

then we have the inequalities:

$$
\begin{align*}
4 m & {\left[\frac{(\sqrt{R}-1)(1-\sqrt{r})}{\sqrt{R}+\sqrt{r}}-h(P \| Q)\right] }  \tag{3.50}\\
& \leq \beta_{f}(r, R)-C_{f}(P \| Q) \\
& \leq 4 M\left[\frac{(\sqrt{R}-1)(1-\sqrt{r})}{\sqrt{R}+\sqrt{r}}-h(P \| Q)\right],
\end{align*}
$$

where $C_{f}(P \| Q), \beta_{f}(r, R)$ and $h(P \| Q)$ as given by (2.1), (2.8) and (1.5) respectively.

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

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

$$
0<r \leq \frac{p_{i}}{q_{i}} \leq R<\infty, \quad \forall i \in\{1,2, \ldots, n\}
$$

then in view of Proposition 3.10, we have

$$
\begin{align*}
4 r^{\frac{2 s-1}{2}} & {\left[\frac{(\sqrt{R}-1)(1-\sqrt{r})}{\sqrt{R}+\sqrt{r}}-h(P \| Q)\right] }  \tag{3.51}\\
& \leq \phi_{s}(r, R)-\Phi_{s}(P \| Q) \\
& \leq 4 R^{\frac{2 s-1}{2}}\left[\frac{(\sqrt{R}-1)(1-\sqrt{r})}{\sqrt{R}+\sqrt{r}}-h(P \| Q)\right], s \geq \frac{1}{2}
\end{align*}
$$

and

$$
\begin{align*}
4 R^{\frac{2 s-1}{2}} & {\left[\frac{(\sqrt{R}-1)(1-\sqrt{r})}{\sqrt{R}+\sqrt{r}}-h(P \| Q)\right] }  \tag{3.52}\\
& \leq \phi_{s}(r, R)-\Phi_{s}(P \| Q) \\
& \leq 4 r^{\frac{2 s-1}{2}}\left[\frac{(\sqrt{R}-1)(1-\sqrt{r})}{\sqrt{R}+\sqrt{r}}-h(P \| Q)\right], \quad s \leq \frac{1}{2}
\end{align*}
$$

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Proof. Let the function $\phi_{s}(u)$ given by (3.29) be defined over $[r, R]$. Defining
$g(u)=u^{3 / 2} \phi_{s}^{\prime \prime}(u)=u^{\frac{2 s-1}{2}}$, obviously we have

$$
\sup _{u \in[r, R]} g(u)= \begin{cases}R^{\frac{2 s-1}{2}}, & s \geq \frac{1}{2}  \tag{3.53}\\ r^{\frac{2 s-1}{2}}, & s \leq \frac{1}{2}\end{cases}
$$

and

$$
\inf _{u \in[r, R]} g(u)= \begin{cases}r^{\frac{2 s-1}{2}}, & s \geq \frac{1}{2}  \tag{3.54}\\ R^{\frac{2 s-1}{2}}, & s \leq \frac{1}{2}\end{cases}
$$

In view of (3.53), (3.54) and Proposition 3.10, we get the proof of the result.
In view of Result 7, we have the following corollary.
Corollary 3.11. Under the conditions of Result 7, we have

$$
\begin{align*}
\frac{8}{\sqrt{R^{3}}} & {\left[\frac{(\sqrt{R}-1)(1-\sqrt{r})}{\sqrt{R}+\sqrt{r}}-h(P \| Q)\right] }  \tag{3.55}\\
& \leq \frac{(R-1)(1-r)}{r R}-\chi^{2}(Q \| P) \\
& \leq \frac{8}{\sqrt{r^{3}}}\left[\frac{(\sqrt{R}-1)(1-\sqrt{r})}{\sqrt{R}+\sqrt{r}}-h(P \| Q)\right],
\end{align*}
$$

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$$
\begin{align*}
& \frac{4}{\sqrt{R}}\left[\frac{(\sqrt{R}-1)(1-\sqrt{r})}{\sqrt{R}+\sqrt{r}}-h(P \| Q)\right]  \tag{3.56}\\
& \quad \leq \frac{(R-1) \ln \frac{1}{r}+(1-r) \ln \frac{1}{R}}{R-r}-K(Q \| P) \\
& \quad \leq \frac{4}{\sqrt{r}}\left[\frac{(\sqrt{R}-1)(1-\sqrt{r})}{\sqrt{R}+\sqrt{r}}-h(P \| Q)\right],
\end{align*}
$$

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

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

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

and

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

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

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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