# SOME NEW INEQUALITIES FOR HERMITE-HADAMARD DIVERGENCE IN INFORMATION THEORY 

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#### Abstract

In this paper we prove some new inequalities for Hermite-Hadamard divergence in Information Theory.


## 1. Introduction

One of the important issues in many applications of Probability Theory is finding an appropriate measure of distance (or difference or discrimination) between two probability distributions. A number of divergence measures for this purpose have been proposed and extensively studied by Jeffreys [1], Kullback and Leibler [2], Rényi [3], Havrda and Charvat [4], Kapur [5], Sharma and Mittal [6], Burbea and Rao [7], Rao [8], Lin [9], Csiszár [10], Ali and Silvey [12], Vajda [13], Shioya and Da-te [40] and others (see for example [5] and the references therein).

These measures have been applied in a variety of fields such as: anthropology [8], genetics [14], finance, economics, and political science [15], [16], [17], biology [18], the analysis of contingency tables [19], approximation of probability distributions [20], [21], signal processing [22], [23] and pattern recognition [24], [25]. A number of these measures of distance are specific cases of $f$-divergence and so further exploration of this concept will have a flow on effect to other measures of distance and to areas in which they are applied.

Let the set $\chi$ and the $\sigma$-finite measure $\mu$ be given and consider the set of all probability densities on $\mu$ to be defined on $\Omega:=\left\{p \mid p: \chi \rightarrow \mathbb{R}, p(x) \geq 0, \int p(x) d \mu(x)=1\right\}$. The Kullback-Leibler divergence [2] is well known among the $\chi$ information divergences. It is defined as:

$$
\begin{equation*}
D_{K L}(p, q):=\int_{\chi} p(x) \log \left[\frac{p(x)}{q(x)}\right] d \mu(x), \quad p, q \in \Omega \tag{1.1}
\end{equation*}
$$

where $\log$ is to base 2 .
In Information Theory and Statistics, various divergences are applied in addition to the Kullback-Leibler divergence. These are the: variation distance $D_{v}$, Hellinger distance $D_{H}$ [1], $\chi^{2}$-divergence $D_{\chi^{2}}$, $\alpha$-divergence $D_{\alpha}$, Bhattacharyya distance $D_{B}[2]$, Harmonic distance $D_{H a}$, Jeffreys distance $D_{J}$ [1], triangular discrimination $D_{\Delta}$ [35], etc... They are defined as follows:

$$
\begin{equation*}
D_{v}(p, q):=\int_{\chi}|p(x)-q(x)| d \mu(x), \quad p, q \in \Omega \tag{1.2}
\end{equation*}
$$

[^0]\[

$$
\begin{gather*}
D_{H}(p, q):=\int_{\chi}|\sqrt{p(x)}-\sqrt{q(x)}| d \mu(x), p, q \in \Omega  \tag{1.3}\\
D_{\chi^{2}}(p, q):=\int_{\chi} p(x)\left[\left(\frac{q(x)}{p(x)}\right)^{2}-1\right] d \mu(x), p, q \in \Omega ;  \tag{1.4}\\
D_{\alpha}(p, q):=\frac{4}{1-\alpha^{2}}\left[1-\int_{\chi}[p(x)]^{\frac{1-\alpha}{2}}[q(x)]^{\frac{1+\alpha}{2}} d \mu(x)\right], p, q \in \Omega ;  \tag{1.5}\\
D_{B}(p, q):=\int_{\chi} \sqrt{p(x) q(x)} d \mu(x), p, q \in \Omega  \tag{1.6}\\
D_{H a}(p, q):=\int_{\chi} \frac{2 p(x) q(x)}{p(x)+q(x)} d \mu(x), p, q \in \Omega  \tag{1.7}\\
D_{J}(p, q):=\int_{\chi}[p(x)-q(x)] \ln \left[\frac{p(x)}{q(x)}\right] d \mu(x), p, q \in \Omega  \tag{1.8}\\
D_{\Delta}(p, q):=\int_{\chi} \frac{[p(x)-q(x)]^{2}}{p(x)+q(x)} d \mu(x), p, q \in \Omega . \tag{1.9}
\end{gather*}
$$
\]

For other divergence measures, see the paper [5] by Kapur or the book on line [6] by Taneja. For a comprehensive collection of preprints available on line, see the RGMIA web site http://rgmia.vu.edu.au/papersinfth.html

Csiszár $f$-divergence is defined as follows [10]

$$
\begin{equation*}
D_{f}(p, q):=\int_{\chi} p(x) f\left[\frac{q(x)}{p(x)}\right] d \mu(x), \quad p, q \in \Omega \tag{1.10}
\end{equation*}
$$

where $f$ is convex on $(0, \infty)$. It is assumed that $f(u)$ is zero and strictly convex at $u=1$. By appropriately defining this convex function, various divergences are derived. All the above distances (1.1)-(1.9), are particular instances of $f$-divergence. There are also many others that are not in this class (see for example [5] or [6]). For the basic properties of $f$-divergence see [7]-[10].

In [11], Lin and Wong (see also [9]) introduced the following divergence

$$
\begin{equation*}
D_{L W}(p, q):=\int_{\chi} p(x) \log \left[\frac{p(x)}{\frac{1}{2} p(x)+\frac{1}{2} q(x)}\right] d \mu(x), \quad p, q \in \Omega . \tag{1.11}
\end{equation*}
$$

This can be represented as follows, using the Kullback-Leibler divergence:

$$
D_{L W}(p, q)=D_{K L}\left(p, \frac{1}{2} p+\frac{1}{2} q\right) .
$$

Lin and Wong have established the following inequalities

$$
\begin{gather*}
D_{L W}(p, q) \leq \frac{1}{2} D_{K L}(p, q)  \tag{1.12}\\
D_{L W}(p, q)+D_{L W}(q, p) \leq D_{v}(p, q) \leq 2  \tag{1.13}\\
D_{L W}(p, q) \leq 1 \tag{1.14}
\end{gather*}
$$

In [45], Shioya and Da-te improved (1.12) - (1.14) by showing that

$$
D_{L W}(p, q) \leq \frac{1}{2} D_{v}(p, q) \leq 1
$$

In the same paper [45], the authors introduced the generalised Lin-Wong $f$ divergence $D_{f}\left(p, \frac{1}{2} p+\frac{1}{2} q\right)$ and the Hermite-Hadamard (HH) divergence

$$
\begin{equation*}
D_{H H}^{f}(p, q):=\int_{\chi} p(x) \frac{\int_{1}^{\frac{q(x)}{p(x)}} f(t) d t}{\frac{q(x)}{p(x)}-1} d \mu(x), \quad p, q \in \Omega \tag{1.15}
\end{equation*}
$$

and, by use of the Hermite-Hadamard inequality for convex functions, proved the following basic inequality

$$
\begin{equation*}
D_{f}\left(p, \frac{1}{2} p+\frac{1}{2} q\right) \leq D_{H H}^{f}(p, q) \leq \frac{1}{2} D_{f}(p, q) \tag{1.16}
\end{equation*}
$$

provided that $f$ is convex and normalised, i.e., $f(1)=0$.
In this paper we point out new inequalites for the $H H$-divergence, which also improve the above result (1.16).

For classical and new results in comparing different kinds of divergence measures, see the papers [1]-[45] where further references are given.

## 2. The Results

In the following, we assume everywhere that the mapping $f:(0, \infty) \rightarrow \mathbb{R}$ is convex and normalised.

The following result holds.
Theorem 1. Let $p, q \in \nless$, then we have the inequality,

$$
\begin{align*}
& D_{f}\left(p, \frac{1}{2} p+\frac{1}{2} q\right)  \tag{2.1}\\
\leq & \lambda D_{f}\left(p, p+\frac{\lambda}{2}(q-p)\right)+(1-\lambda) D_{f}\left(p, \frac{p+q}{2}+\frac{\lambda}{2}(q-p)\right) \\
\leq & D_{H H}^{f}(p, q) \leq \frac{1}{2}\left[D_{f}(p,(1-\lambda) p+\lambda q)+(1-\lambda) D_{f}(p, q)\right] \\
\leq & \frac{1}{2} D_{f}(p, q)
\end{align*}
$$

for all $\lambda \in[0,1]$.
Proof. First, the following refinement of the Hermite-Hadamard inequality is proved.

$$
\begin{align*}
& f\left(\frac{a+b}{2}\right)  \tag{2.2}\\
\leq & \lambda f\left(a+\lambda \cdot \frac{b-a}{2}\right)+(1-\lambda) f\left(\frac{a+b}{2}+\lambda \cdot \frac{b-a}{2}\right) \\
\leq & \frac{1}{b-a} \int_{a}^{b} f(u) d u \leq \frac{1}{2}[f((1-\lambda) a+\lambda b)+\lambda f(a)+(1-\lambda) f(b)] \\
\leq & \frac{f(a)+f(b)}{2}
\end{align*}
$$

for all $\lambda \in[0,1]$.

Applying the Hermite-Hadamard inequality on each subinterval $[a,(1-\lambda) a+\lambda b]$, $[(1-\lambda) a+\lambda b, b]$, we have,

$$
\begin{aligned}
& f\left(\frac{a+(1-\lambda) a+\lambda b}{2}\right) \times[(1-\lambda) a+\lambda b-a] \\
\leq & \int_{a}^{(1-\lambda) a+\lambda b} f(u) d u \\
\leq & \frac{f((1-\lambda) a+\lambda b)+f(a)}{2} \times[(1-\lambda) a+\lambda b-a]
\end{aligned}
$$

and

$$
\begin{aligned}
& f\left(\frac{(1-\lambda) a+\lambda b+b}{2}\right) \times[b-(1-\lambda) a-\lambda b] \\
\leq & \int_{(1-\lambda) a+\lambda b}^{b} f(u) d u \\
\leq & \frac{f(b)+f((1-\lambda) a+\lambda b)}{2} \times[b-(1-\lambda) a-\lambda b],
\end{aligned}
$$

which are clearly equivalent to

$$
\begin{align*}
\lambda f\left(a+\lambda \cdot \frac{b-a}{2}\right) & \leq \frac{1}{b-a} \int_{a}^{(1-\lambda) a+\lambda b} f(u) d u  \tag{2.3}\\
& \leq \frac{\lambda f((1-\lambda) a+\lambda b)+\lambda f(a)}{2}
\end{align*}
$$

and

$$
\begin{align*}
& (1-\lambda) f\left(\frac{a+b}{2}+\lambda \cdot \frac{b-a}{2}\right)  \tag{2.4}\\
\leq & \frac{1}{b-a} \int_{(1-\lambda) a+\lambda b}^{b} f(u) d u \\
\leq & \frac{(1-\lambda) f(b)+(1-\lambda) f((1-\lambda) a+\lambda b)}{2}
\end{align*}
$$

respectively.
Summing (2.3) and (2.4), we obtain the second and first inequality in (2.2).
By the convexity property, we obtain

$$
\begin{aligned}
& \lambda f\left(a+\lambda \cdot \frac{b-a}{2}\right)+(1-\lambda) f\left(\frac{a+b}{2}+\lambda \cdot \frac{b-a}{2}\right) \\
\geq & f\left[\lambda\left(a+\lambda \cdot \frac{b-a}{2}\right)+(1-\lambda)\left(\frac{a+b}{2}+\lambda \cdot \frac{b-a}{2}\right)\right] \\
= & f\left(\frac{a+b}{2}\right)
\end{aligned}
$$

and the first inequality in (2.1) is proved.
The latter inequality is obvious by the convexity property of $f$.

Now, if we choose $a=1$ and $b=\frac{q(x)}{p(x)}, x \in \chi$, in (2.2) and multiply by $p(x) \geq 0$, $x \in \chi$, we get

$$
\begin{aligned}
& p(x) f\left(\frac{p(x)+q(x)}{2 p(x)}\right) \\
\leq & \lambda p(x) f\left(\frac{p(x)+\lambda(q(x)-p(x))}{2 p(x)}\right) \\
& +(1-\lambda) p(x) f\left(\frac{p(x)+q(x)}{2 p(x)}+\frac{\lambda(q(x)-p(x))}{2 p(x)}\right) \\
\leq & \frac{p^{2}(x)}{q(x)-p(x)} \int_{1}^{\frac{q(x)}{p(x)}} f(u) d u \\
\leq & \frac{1}{2}\left[f\left(\frac{(1-\lambda) p(x)+\lambda q(x)}{p(x)}\right) p(x)+\lambda p(x) f(1)+(1-\lambda) p(x) f\left(\frac{q(x)}{p(x)}\right)\right] \\
\leq & \frac{p(x) f(1)+p(x) f\left(\frac{q(x)}{p(x)}\right)}{2} .
\end{aligned}
$$

Integrating on $\chi$ and taking into account the definition of $f$-divergence (1.10) and the Hermite-Hadamard divergence (1.15), we obtain (2.1).

Remark 1. If $\lambda=0$ or $\lambda=1$, then by (2.1), we obtain the inequality (1.16).
Corollary 1. Let $p, q \in \Omega$, then we have the inequality,

$$
\begin{align*}
D_{f}\left(p, \frac{p+q}{2}\right) & \leq \frac{1}{2}\left[D_{f}\left(p, \frac{3 p+q}{4}\right)+D_{f}\left(p, \frac{p+3 q}{4}\right)\right]  \tag{2.5}\\
& \leq D_{H H}^{f}(p, q) \leq \frac{1}{2}\left[D_{f}\left(p, \frac{p+q}{2}\right)+\frac{1}{2} D_{f}(p, q)\right] \\
& \leq \frac{1}{2} D_{f}(p, q)
\end{align*}
$$

which is obtained by taking $\lambda=\frac{1}{2}$ in (2.1).
Remark 2. If we replace $\lambda$ by $(1-\lambda)$ in (2.1), we have,

$$
\begin{align*}
& D_{f}\left(p, \frac{p+q}{2}\right)  \tag{2.6}\\
\leq & (1-\lambda) D_{f}\left(p, \frac{p+q}{2}+\lambda(p-q)\right)+\lambda D_{f}\left(p, q+\lambda \frac{p-q}{2}\right) \\
\leq & D_{H H}^{f}(p, q) \leq \frac{1}{2}\left[D_{f}(p, \lambda p+(1-\lambda) q)+\lambda D_{f}(p, q)\right] \\
\leq & \frac{1}{2} D_{f}(p, q) .
\end{align*}
$$

Now, if we add (2.1) and (2.6) and divide by 2, we can state the following corollary.

Corollary 2. Let $p, q \in \Omega$, then we have the inequality,

$$
\begin{align*}
& D_{f}\left(p, \frac{p+q}{2}\right)  \tag{2.7}\\
\leq & \lambda\left[D_{f}\left(p, p+\frac{\lambda}{2}(q-p)\right)+D_{f}\left(p, q+\frac{\lambda}{2}(p-q)\right)\right] \\
& +(1-\lambda)\left[D_{f}\left(p, \frac{p+q}{2}+\frac{\lambda}{2}(q-p)\right)+D_{f}\left(p, \frac{p+q}{2}+\frac{1}{2}(p-q)\right)\right] \\
\leq & D_{H H}^{f}(p, q) \\
\leq & \frac{1}{4}\left[D_{f}(p,(1-\lambda) p+\lambda q)+D_{f}(p, \lambda p+(1-\lambda) q)+D_{f}(p, q)\right] \\
\leq & \frac{1}{2} D_{f}(p, q)
\end{align*}
$$

for all $\lambda \in[0,1]$.
We also define the divergence.

$$
\begin{align*}
H_{f}(p, q ; t) & : \quad \int_{\chi} p(x) f\left[\frac{t q(x)+(1-t) p(x)}{p(x)}\right] d \mu(x)  \tag{2.8}\\
& =D_{f}(p, t q+(1-t) p)
\end{align*}
$$

Theorem 2. Let $p, q \in \Omega$, then,
(i) $H_{f}(p, q ; \cdot)$ is convex on $[0,1]$;
(ii) We have the bounds

$$
\begin{gather*}
\inf _{t \in[0,1]} H_{f}(p, q ; t)=H_{f}(p, q ; 0)=0,  \tag{2.9}\\
\sup _{t \in[0,1]} H_{f}(p, q ; t)=H_{f}(p, q ; 1)=D_{f}(p, q), \tag{2.10}
\end{gather*}
$$

and the inequality

$$
\begin{equation*}
H_{f}(p, q ; t) \leq t D_{f}(p, q) \text { for all } t \in[0,1] \tag{2.11}
\end{equation*}
$$

(iii) The mapping $H_{f}(p, q ; \cdot)$ is monotonic nondecreasing on $[0,1]$.

Proof. (i) Let $t_{1}, t_{2} \in[0,1]$ and $\alpha, \beta \in[0,1]$ with $\alpha+\beta=1$, then,

$$
\begin{aligned}
& H_{f}\left(p, q ; \alpha t_{1}+\beta t_{2}\right) \\
= & \int_{\chi} p(x) f\left[\frac{\left(\alpha t_{1}+\beta t_{2}\right) q(x)+\left(1-\alpha t_{1}-\beta t_{2}\right) p(x)}{q(x)}\right] d \mu(x) \\
= & \int_{\chi} p(x) f\left[\alpha \cdot \frac{\left[t_{1} q(x)+\left(1-t_{1}\right) p(x)\right]}{q(x)}+\beta \cdot \frac{\left[t_{2} q(x)+\left(1-t_{2}\right) p(x)\right]}{q(x)}\right] d \mu(x) \\
\leq & \alpha \cdot \int_{\chi} p(x) f\left[\frac{t_{1} q(x)+\left(1-t_{1}\right) p(x)}{q(x)}\right] d \mu(x) \\
& +\beta \cdot \int_{\chi} p(x) f\left[\frac{t_{2} q(x)+\left(1-t_{2}\right) p(x)}{q(x)}\right] d \mu(x) \\
= & \alpha H_{f}\left(p, q, t_{1}\right)+\beta H_{f}\left(p, q, t_{2}\right)
\end{aligned}
$$

and convexity is proved.
(ii) Using Jensen's inequality, we have:

$$
\begin{aligned}
H_{f}(p, q, t) & \geq f\left[\int_{\chi} p(x)\left[\frac{t q(x)+(1-t) p(x)}{q(x)}\right] d \mu(x)\right] \\
& =f\left[t \int_{\chi} q(x) d \mu(x)+(1-t) \int_{\chi} p(x) d \mu(x)\right] \\
& =f(1)=0=H_{f}(p, q, 0)
\end{aligned}
$$

Also, by convexity of $f$, we have,

$$
\begin{aligned}
H_{f}(p, q, t) & \leq \int_{\chi} p(x)\left[t f\left(\frac{q(x)}{p(x)}\right)+(1-t) f(1)\right] d \mu(x) \\
& \leq t \int_{\chi} p(x) f\left(\frac{q(x)}{p(x)}\right) d \mu(x)+(1-t) f(1) \int_{\chi} p(x) d \mu(x) \\
& =t D_{f}(p, q)
\end{aligned}
$$

and the statement (ii) is proved.
(iii) Let $t_{1}, t_{2} \in[0,1]$ with $t_{2}>t_{1}$. As $H_{f}(p, q ; \cdot)$ is convex, then

$$
\frac{H_{f}\left(p, q, t_{2}\right)-H_{f}\left(p, q, t_{1}\right)}{t_{2}-t_{1}} \geq \frac{H_{f}\left(p, q, t_{1}\right)-H_{f}(p, q, 0)}{t_{1}-0}
$$

and as

$$
H_{f}\left(p, q, t_{1}\right) \geq H_{f}(p, q, 0)=0
$$

we deduce that $H_{f}\left(p, q, t_{1}\right) \leq H_{f}\left(p, q, t_{2}\right)$, which proves the monotonicity of $H_{f}(p, q, \cdot)$.

Remark 3. If we write (2.11) in terms of $1-t$ rather than $t$, we obtain

$$
\begin{equation*}
H_{f}(p, q, 1-t) \leq(1-t) D_{f}(p, q), \quad t \in[0,1] \tag{2.12}
\end{equation*}
$$

Adding (2.11) and (2.12), we get,

$$
\begin{equation*}
H_{f}(p, q, t)+H_{f}(p, q, 1-t) \leq D_{f}(p, q) \tag{2.13}
\end{equation*}
$$

for all $t \in[0,1]$.
Remark 4. For $t \in\left[\frac{1}{2}, 1\right]$, we have the inequality,

$$
\begin{equation*}
D_{f}\left(p, \frac{1}{2} p+\frac{1}{2} q\right) \leq D_{f}(p, t q+(1-t) p) \leq t D_{f}(p, q) \tag{2.14}
\end{equation*}
$$

which is similar to (1.13).
We can also define the divergence,

$$
\begin{equation*}
F_{f}(p, q ; t):=\int_{\chi} \int_{\chi} p(x) p(y) f\left[t \cdot \frac{q(x)}{p(x)}+(1-t) \cdot \frac{q(y)}{p(y)}\right] d \mu(x) d \mu(y) \tag{2.15}
\end{equation*}
$$

where $p, q \in \Omega$ and $t \in[0,1]$.
The properties of this mapping are embodied in the following theorem.
Theorem 3. Let $p, q \in \Omega$, then,
(i) $F_{f}(p, q ; \cdot)$ is symmetrical about $\frac{1}{2}$, that is,

$$
\begin{equation*}
F_{f}(p, q ; t)=F_{f}(p, q ; 1-t) \quad \text { for all } t \in[0,1] . \tag{2.16}
\end{equation*}
$$

(ii) $F$ is convex on $[0,1]$;
(iii) We have the bounds:

$$
\begin{align*}
& \sup _{t \in[0,1]} F_{f}(p, q ; t)=F_{f}(p, q ; 0)=F_{f}(p, q ; 1)=D_{f}(p, q)  \tag{2.17}\\
& \inf _{t \in[0,1]} F_{f}(p, q ; t)=F_{f}\left(p, q ; \frac{1}{2}\right) \\
= & \int_{\chi} \int_{\chi} p(x) p(y) f\left[\frac{q(x) p(y)+p(x) q(y)}{2 p(x) q(y)}\right] d \mu(x) d \mu(y) \\
\geq & 0
\end{align*}
$$

(iv) $F_{f}(p, q ; \cdot)$ is nondecreasing on $\left[0, \frac{1}{2}\right]$ and nonincreasing on $\left[\frac{1}{2}, 1\right]$;
(v) We have the inequality:

$$
\begin{equation*}
F_{f}(p, q ; t) \geq \max \left\{H_{f}(p, q ; t) ; H_{f}(p, q ; 1-t)\right\} \quad \text { for all } t \in[0,1] \tag{2.19}
\end{equation*}
$$

Proof.
(i) Is obvious.
(ii) Follows by the convexity of $f$ in a similar way to that in the proof of Theorem 2.
(iii) For all $x, y \in \chi$ we have:

$$
f\left[t \cdot \frac{q(x)}{p(x)}+(1-t) \cdot \frac{q(y)}{p(y)}\right] \leq t \cdot f\left(\frac{q(x)}{p(x)}\right)+(1-t) \cdot f\left(\frac{q(y)}{p(y)}\right)
$$

for any $t \in[0,1]$.
Multiplying by $p(x) p(y) \geq 0$ and integrating over $\chi^{2}$, we write,

$$
\begin{aligned}
F_{f}(p, q ; t) \leq & \int_{\chi} \int_{\chi} p(x) p(y)\left[t \cdot f\left(\frac{q(x)}{p(x)}\right)+(1-t) \cdot f\left(\frac{q(y)}{p(y)}\right)\right] d \mu(x) d \mu(y) \\
= & t \int_{\chi} p(y) d \mu(y) \int_{\chi} p(x) f\left(\frac{q(x)}{p(x)}\right) d \mu(x) \\
& +(1-t) \int_{\chi} d \mu(x) \int_{\chi} p(y) f\left(\frac{q(y)}{p(y)}\right) d \mu(y) \\
= & t \cdot D_{f}(p, q)+(1-t) \cdot D_{f}(p, q)=D_{f}(p, q) \\
= & F_{f}(p, q ; 0)=F_{f}(p, q ; 1)
\end{aligned}
$$

and the bound (2.17) is proved.
Since $f$ is convex, then for all $t \in[0,1]$ and $x, y \in \chi$, we have

$$
\begin{aligned}
& \frac{1}{2}\left\{f\left[t \cdot \frac{q(x)}{p(x)}+(1-t) \cdot \frac{q(y)}{p(y)}\right]+f\left[(1-t) \cdot \frac{q(x)}{p(x)}+t \cdot \frac{q(y)}{p(y)}\right]\right\} \\
\geq & f\left[\frac{1}{2}\left(\frac{q(x)}{p(x)}+\frac{q(y)}{p(y)}\right)\right] .
\end{aligned}
$$

Multiplying by $p(x) p(y) \geq 0$ and integrating over $\chi^{2}$, we have,

$$
\begin{aligned}
& \frac{1}{2}\left[F_{f}(p, q ; t)+F_{f}(p, q ; 1-t)\right] \\
\geq & \int_{\chi} \int_{\chi} p(x) p(y) f\left[\frac{1}{2}\left(\frac{q(x)}{p(x)}+\frac{q(y)}{p(y)}\right)\right] d \mu(x) d \mu(y)
\end{aligned}
$$

and the first part of (2.18) is proved.
Using Jensen's integral inequality, we may write:

$$
\begin{aligned}
& \int_{\chi} \int_{\chi} f\left[\frac{1}{2}\left(\frac{q(x) p(y)+p(x) q(y)}{p(x) q(y)}\right)\right] p(x) p(y) d \mu(x) d \mu(y) \\
\geq & f\left[\int_{\chi} \int_{\chi} \frac{1}{2}\left(\frac{q(x) p(y)+p(x) q(y)}{p(x) q(y)}\right) p(x) p(y) d \mu(x) d \mu(y)\right] \\
= & f\left[\frac{1}{2}\left[\int_{\chi} p(x) d \mu(x) \int_{\chi} p(y) d \mu(y)+\int_{\chi} q(x) d \mu(x) \int_{\chi} q(y) d \mu(y)\right]\right] \\
= & f(1)=0
\end{aligned}
$$

and the second part of (2.18) is proved.
(iv) The mapping $F_{f}(p, q ; \cdot)$ being convex on $[0,1]$, we may write, for $1 \geq t_{2}>$ $t_{1} \geq \frac{1}{2}$, that,

$$
\frac{F_{f}\left(p, q ; t_{2}\right)-F_{f}\left(p, q ; t_{1}\right)}{t_{2}-t_{1}} \geq \frac{F_{f}\left(p, q ; t_{1}\right)-F_{f}\left(p, q ; \frac{1}{2}\right)}{t_{1}-\frac{1}{2}}
$$

and as

$$
F_{f}\left(p, q ; t_{1}\right) \geq F_{f}\left(p, q ; \frac{1}{2}\right), \quad t_{1} \geq \frac{1}{2}
$$

we deduce that $F_{f}\left(p, q ; t_{2}\right) \geq F_{f}\left(p, q ; t_{1}\right)$, i.e., the mapping $F_{f}(p, q ; \cdot)$ is monotonically nondecreasing on $\left[0, \frac{1}{2}\right]$.
Similarly, we can prove that $F_{f}(p, q ; \cdot)$ is monotonically nonincreasing on [ $0, \frac{1}{2}$ ], and the statement (iv) is proved.
(v) Using Jensen's integral inequality, we have,

$$
\begin{aligned}
& \int_{\chi} p(y) f\left[t \cdot \frac{q(x)}{p(x)}+(1-t) \cdot \frac{q(y)}{p(y)}\right] d \mu(y) \\
\geq & f\left[\int_{\chi} p(y)\left[t \cdot \frac{q(x)}{p(x)}+(1-t) \cdot \frac{q(y)}{p(y)}\right] d \mu(y)\right] \\
= & f\left[t \cdot \frac{q(x)}{p(x)} \int_{\chi} p(y) d \mu(y)+(1-t) \cdot \int_{\chi} q(y) d \mu(y)\right] \\
= & f\left[t \cdot \frac{q(x)}{p(x)}+(1-t)\right] .
\end{aligned}
$$

Multiplying by $p(x) \geq 0$ and integrating over $\chi$, we have,

$$
\begin{aligned}
F_{f}(p, q ; t) & \geq \int_{\chi} p(x) f\left[t \cdot \frac{q(x)}{p(x)}+(1-t)\right] d \mu(x) \\
& =H_{f}(p, q ; t)
\end{aligned}
$$

for all $t \in[0,1]$.
Now, as

$$
F_{f}(p, q ; 1-t) \geq H_{f}(p, q ; 1-t)
$$

and $F_{f}(p, q ; t)=F_{f}(p, q ; 1-t)$ for all $t \in[0,1]$, the inequality (2.19) is completely proved.

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